3) Seemingly Unrelated Regressions (SUR)

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System of Equations

$$\begin{aligned} \textit{housing} &= \beta_{10} + \beta_{11} \textit{houseprc} + \beta_{12} \textit{foodprc} + \beta_{13} \textit{clothprc} + \beta_{14} \textit{income} \\ &+ \beta_{15} \textit{size} + \beta_{16} \textit{age} + \textit{u}_1. \end{aligned}$$

$$food = \beta_{20} + \beta_{21}houseprc + \beta_{22}foodprc + \beta_{23}clothprc + \beta_{24}income$$
$$+ \beta_{25}size + \beta_{26}age + u_2.$$

clothing =
$$\beta_{30} + \beta_{31}$$
houseprc + β_{32} foodprc + β_{33} clothprc + β_{34} income + β_{35} size + β_{36} age + u_3 .

$$E(u_g|x_1, x_2, ..., x_G) = 0$$

 $g = 1, ..., G$

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Generalized Least Squares (GLS)

$$E(\epsilon \epsilon' | X) = \sigma^2 V(X)$$
 $(n \times n)$
 $V^{-1} = C'C$
 $Cy = CX\beta + C\epsilon$
 $\hat{\beta}_{GLS} = (X'V^{-1}X)^{-1}X'V^{-1}y$

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GLS - Two Equations

$$y_g = X_g \beta_g + u_g, \quad g = 1, 2$$

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix} \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} + \begin{pmatrix} u_1 \\ u_2 \end{pmatrix}$$

$$\Sigma \otimes I_N = \begin{bmatrix} \sigma_{11}I_N & \sigma_{12}I_N \\ \sigma_{21}I_N & \sigma_{22}I_N \end{bmatrix}$$

$$\hat{\beta}_{GLS} = \begin{bmatrix} \sigma_{11} X_1' X_1 & \sigma_{12} X_1' X_2 \\ \sigma_{21} X_2' X_1 & \sigma_{22} X_2' X_2 \end{bmatrix}^{-1} \begin{bmatrix} \sigma_{11} X_1' y_1 + \sigma_{12} X_1' y_2 \\ \sigma_{21} X_2' y_1 + \sigma_{22} X_2' y_2 \end{bmatrix}$$

Seemingly Unrelated Regressions (SUR)

$$\begin{bmatrix} y_{i1} \\ \vdots \\ y_{iG} \end{bmatrix} = \begin{bmatrix} x'_{i1} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & x'_{iG} \end{bmatrix} \begin{bmatrix} \beta_{i1} \\ \vdots \\ \beta_{iG} \end{bmatrix} + \begin{bmatrix} u_{i1} \\ \vdots \\ u_{iG} \end{bmatrix}$$

$$\hat{\beta}_{GLS} = \{X'(\Sigma^{-1} \otimes I_N)X\}^{-1}\{X'(\Sigma^{-1} \otimes I_N)y\}$$

$$Var(\hat{eta}) = \{X'(\Sigma^{-1} \otimes I_N)X\}^{-1}$$

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SUR Estimator = **FGLS Estimator**

- 1) Estimate each equation by OLS
 - 2) Estimate Σ , using:

$$\hat{u}_j = y_j - X_j \hat{eta}_j$$
 and $\hat{\sigma}_{jj'} = \hat{u}_j' \hat{u}_{j'}/N$

3) Use $\hat{\Sigma}$ to obtain $\hat{\beta}_{FGLS}$

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SUR Model for Hourly Wages and Hourly Benefits

12	=	.32	

Explanatory Variables	hrearn	hrbens
educ	.459 (.069)	.077 (.008)
exper	076 (.057)	.023 (.007)
$exper^2$.0040 (.0012)	0005 (.0001)
tenure	.110 (.084)	.054 (.010)
tenure ²	0051 (.0033)	0012 (.0004)
union	.808 (.408)	.366 (.049)
south	457 (.552)	023 (.066)
nrtheast	-1.151 (0.606)	057 (.072)
nrthcen	636 (.556)	038 (.066)
married	.642 (.418)	.058
white	1.141 (0.612)	.090 (.073)
male	1.785 (0.398)	(.050) .090 (.073) .268 (.048)
inter cept	-2.632 (1.228)	890 (.147)

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Medical Expenditure Panel Survey (MEPS)

- Medicare-eligible population
- \bullet Aged > 65 years
- Medicare does not cover all medical expenses
- People usually buy private insurance

Idrugexp: log of expenditure on prescribed drugs **Idrugexp**: log of expenditure on all categories of

medical services other than drugs

actlim: activity limitation

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use mus05surdata.dta

summarize ldrugexp ltotothr age age2 educyr /// actlim totchr medicaid private

Variable	Obs	Mean	Std. Dev.	Min	Max
ldrugexp ltotothr age age2 educyr	3,285 3,350 3,384 3,384 3,384	6.936533 7.537196 74.38475 5573.898 11.29108	1.300312 1.61298 6.388984 961.357 3.7758	1.386294 1.098612 65 4225	10.33773 11.71892 90 8100 17
actlim totchr medicaid private	3,384 3,384 3,384 3,384	.3454492 1.954492 .161643 .5156619	.4755848 1.326529 .3681774 .4998285	0 0 0	1 8 1 1

reg ldrugexp age age2 actlim totchr medicaid private

Source	SS	df	MS		er of obs		3,285
Model Residual	1260.35424 4292.27045	6 3,278	210.05904 1.30941746	Prob R-sq	3278) > F [uared R-squared	= = = i =	160.42 0.0000 0.2270 0.2256
Total	5552.62469	3,284	1.69081141		MSE	=	1.1443
ldrugexp	Coef.	Std. Err.	t	P> t	[95% (Conf.	Interval]
age age2 actlim totchr medicaid private _cons	.27641480018315 .357446 .4035182 .0893386 .0775393 -4.402228	.0798228 .0005306 .0468032 .0162256 .0600878 .0438092 2.986652	-3.45 7.64 24.87 1.49 1.77	0.001 0.001 0.000 0.000 0.137 0.077	.11990 00287 .26567 .37170 02847 00835	718 795 949 748	.4329224 0007911 .4492125 .4353316 .2071521 .1634355 1.453664

reg Itotothr age age2 educyr actlim totchr private

Source	SS	df	MS		er of obs		3,350
Model Residual	1342.02084 7371.09274	6 3,3 4 3	223.67014 2.20 4 93352	Prob	3343) > F puared R-squared	= = = ! =	101.44 0.0000 0.1540 0.1525
Total	8713.11358	3,349	2.60170606		MSE	=	1.4849
ltotothr	Coef.	Std. Err.	t	P> t	[95% C	onf.	Interval]
age age2 educyr actlim totchr private _cons	.3173817 0020875 .0650207 .7421208 .295988 .258998 -6.1414	.1026155 .0006823 .0072825 .0601413 .0206706 .0546362 3.838185	3.09 -3.06 8.93 12.34 14.32 4.74 -1.60	0.002 0.002 0.000 0.000 0.000 0.000 0.0110	.11618 00342 .05074 .62420 .25545 .15187	252 21 32 37 41	.5185772 0007498 .0792993 .8600384 .3365164 .3661218

$$r_{12} = \hat{\sigma}_{12} / \sqrt{\hat{\sigma}_{11}\hat{\sigma}_{22}} = 0.17$$

sureg (ldrugexp age age2 actlim totchr medicaid private) /// (Itotothr age age2 educyr actlim totchr private), corr

Equation	Obs	Parms	RMSE	"R-sq"	chi2	P
ldrugexp	3,251	6	1.133657	0.2284	962.07	0.0000
ltotothr	3,251	6	1.491159	0.1491	567.91	

```
Correlation matrix of residuals:

ldrugexp ltotothr
ldrugexp 1.0000
ltotothr 0.1741 1.0000
```

Breusch-Pagan test of independence: chi2(1) = 98.590, Pr = 0.0000

$$Nr_{12}^2 = 3251 \times 0.1741^2 = 98.54$$

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SUR Results

	Coef.	Std. Err.	Z	P> z	[95% Conf.	Interval]
ldrugexp						
age	.2630418	.0795316	3.31	0.001	.1071627	.4189209
age2	0017428	.0005287	-3.30	0.001	002779	0007066
actlim	.3546589	.046617	7.61	0.000	.2632912	.4460266
totchr	.4005159	.0161432	24.81	0.000	.3688757	.432156
medicaid	.1067772	.0592275	1.80	0.071	0093065	.2228608
private	.0810116	.0435596	1.86	0.063	0043636	.1663867
_cons	-3.891259	2.975898	-1.31	0.191	-9.723911	1.941394
ltotothr						
age	.2927827	.1046145	2.80	0.005	.087742	.4978234
age2	0019247	.0006955	-2.77	0.006	0032878	0005617
educyr	.0652702	.00732	8.92	0.000	.0509233	.0796172
actlim	.7386912	.0608764	12.13	0.000	.6193756	.8580068
totchr	.2873668	.0211713	13.57	0.000	.2458719	.3288618
private	.2689068	.055683	4.83	0.000	.1597701	.3780434
cons	-5.198327	3.914053	-1.33	0.184	-12.86973	2.473077

bootstrap, reps(400) seed(10101) nodots: sureg ///

(ldrugexp age age2 actlim totchr medicaid private) /// (ltotothr age age2 educyr actlim totchr private)

	Observed	Bootstrap			Normal	-based
	Coef.	Std. Err.	Z	P> z	[95% Conf.	<pre>Interval]</pre>
ldrugexp						
age	.2630418	.0743481	3.54	0.000	.1173222	.4087614
age2	0017428	.0004929	-3.54	0.000	0027089	0007766
actlim	. 3546589	.0462869	7.66	0.000	.2639382	.4453795
totchr	. 4005159	.0169809	23.59	0.000	.3672339	. 4337979
medicaid	.1067772	.0642814	1.66	0.097	019212	.2327664
private	.0810116	.044791	1.81	0.071	0067771	.1688002
_cons	-3.891259	2.794579	-1.39	0.164	-9.368532	1.586015
ltotothr						
age	.2927827	.1062298	2.76	0.006	.0845762	.5009892
age2	0019247	.0007048	-2.73	0.006	0033061	0005434
educyr	.0652702	.0075052	8.70	0.000	.0505602	.0799802
actlim	.7386912	.0619353	11.93	0.000	.6173003	.8600821
totchr	.2873668	.0202824	14.17	0.000	.247614	.3271196
private	.2689068	.0548669	4.90	0.000	.1613696	.376444
_cons	-5.198327	3.991338	-1.30	0.193	-13.02121	2.624553

test age age2

test [ldrugexp]private = [ltotothr]private

(1) [ldrugexp]private - [ltotothr]private = 0

```
chi2(1) = 8.35
Prob > chi2 = 0.0038
```

constraint 1 [ldrugexp]private = [ltotothr]private

sureg (ldrugexp age age2 actlim totchr medicaid private) /// (ltotothr age age2 educyr actlim totchr private), constraints(1)

(1) [ldruge	exp]private -	[ltotothr]p	rivate =	0		
	Coef.	Std. Err.	z	P> z	[95% Conf	. Interval]
ldrugexp						
age	.2707053	.0795434	3.40	0.001	.1148031	.4266076
age2	0017907	.0005288	-3.39	0.001	0028271	0007543
actlim	.3575386	.0466396	7.67	0.000	.2661268	.4489505
totchr	.3997819	.0161527	24.75	0.000	.3681233	.4314405
medicaid	.1473961	.0575962	2.56	0.010	.0345096	.2602827
private	.1482936	.0368364	4.03	0.000	.0760955	.2204917
_cons	-4.235088	2.975613	-1.42	0.155	-10.06718	1.597006
totothr						
age	.2780287	.1045298	2.66	0.008	.073154	.4829034
age2	0018298	.0006949	-2.63	0.008	0031919	0004677
educyr	.0703523	.0071112	9.89	0.000	.0564147	.0842899
actlim	.7276336	.0607791	11.97	0.000	.6085088	.8467584
totchr	.2874639	.0211794	13.57	0.000	. 245953	.3289747
private	.1482936	.0368364	4.03	0.000	.0760955	.2204917
cons	-4.62162	3.910453	-1.18	0.237	-12.28597	3.042727

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