

DEMONSTRATING MULTI-COLINEARITY WITH THE CONSUMPTION FUNCTION FOR THE U.S.

Here's an example of multi-collinearity using the data from assignment 3 which also appeared on the second exam.

```
. use assg3
```

```
. summarize
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Variable	Obs	Mean	Std. Dev.	Min	Max
num	20	11.5	5.91608	2	21
year	20	1977.5	5.91608	1968	1987
gnp	20	2347.765	1188.909	892.7	4486.2
cnsmp	20	1506.075	792.4492	552.5	2966
invest	20	373.2	204.1194	15.2	716.4
intrate	20	9.6775	3.546692	5.25	18.87
deflator	20	74.795	27.74539	39.3	118.9
rgnp	20	29.81999	4.608627	22.71501	37.73087
rcnsmpt	20	19.00449	3.291805	14.05852	24.94533
rinvest	20	4.661046	1.32795	.3707317	6.149861
rintrate	20	.1370528	.0411115	.0691337	.1994715
d80	20	.4	.5026247	0	1
d80gnp	20	1443	1851.497	0	4486.2
d80intr	20	4.9155	6.584569	0	18.87
d80invst	20	229.795	296.4755	0	716.4
d8rgnp	20	13.68699	17.26466	0	37.73087
d8rintr	20	.048513	.0676758	0	.1994715

Here's the basic regression relating consumption to gnp and interest rates

```
. regress cnsmp gnp intrate
```

Source	SS	df	MS	Number of obs = 20
Model	11924985.5	2	5962492.77	F(2, 17) = 15470.24
Residual	6552.08842	17	385.416966	Prob > F = 0.0000
Total	11931537.6	19	627975.664	R-square = 0.9995 Adj R-square = 0.9994 Root MSE = 19.632

cnsmp	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
gnp	.6752645	.0042736	158.008	0.000	.6662479 .684281
intrate	-6.621716	1.432583	-4.622	0.000	-9.644202 -3.599229
_cons	-15.20564	13.54168	-1.123	0.277	-43.77609 13.36481

Now we want to see whether there was a shift in the consumption function during the 1980s so we introduce a dummy variable d80=1 if year>1979, 0 otherwise.

```
. regress cnsmp gnp intrate d80
```

Source	SS	df	MS	Number of obs = 20
				F(3, 16) = 15491.70

Model	11927431.4	3	3975810.45	Prob > F	=	0.0000
Residual	4106.26104	16	256.641315	R-square	=	0.9997
Total	11931537.6	19	627975.664	Adj R-square	=	0.9996

cnsmppt	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
gnp	.6568772	.006902	95.172	0.000	.6422457 .6715088
intrate	-8.72962	1.353815	-6.448	0.000	-11.59958 -5.859661
d80	56.75664	18.38515	3.087	0.007	17.78188 95.73141
_cons	25.6599	17.24354	1.488	0.156	-10.89478 62.21458

No problems to this point. The coefficient for d80 is significantly different from zero so it looks like there was an upward shift in the consumption function in the 1980s. Now we want to see if the marginal propensity to consume was different in the 80s than it was previously so we form d80gnp=d80*gnp and add it to the regression.

. regress cnsmppt gnp intrate d80 d80gnp

Source	SS	df	MS	Number of obs	=	20
Model	11927705.8	4	2981926.46	F(4, 15)	=	11673.12
Residual	3831.78654	15	255.452436	Prob > F	=	0.0000
Total	11931537.6	19	627975.664	R-square	=	0.9997

cnsmppt	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
gnp	.6485717	.010565	61.389	0.000	.626053 .6710903
intrate	-6.99172	2.152975	-3.247	0.005	-11.58068 -2.402762
d80	-13.35875	70.08501	-0.191	0.851	-162.7414 136.0239
d80gnp	.0221732	.0213911	1.037	0.316	-.0234208 .0677673
_cons	24.39106	17.24705	1.414	0.178	-12.37015 61.15228

Now we have a problem. The coefficient on the new variable, d80gnp, is not significant and the d80 coefficient is now insignificant as well. Notice that R^2 and Adj R^2 did not change from the previous regression to this one. This is a problem of multi-collinearity. If we do an F test to test

$$H_0 = b_{c,d80,i,gnp,d80gnp} = b_{c,d80gnp,i,gnp,d80} = 0$$

by comparing this regression, as the unrestricted regression to the base regression, as the restricted regression, we get:

$$F_{\text{sample}} = (.9997 - .9995) / (1 - .9997) * (20 - 5) / (5 - 3) = 5.0 \text{ which is } > F_{\text{critical}} = F_{2,15} = 3.68$$

So we can clearly reject H_0 .

Notice the steps necessary to establish the presence of the multi-collinearity problem.

1. The regression with one of the variables added (in this case d80) yields a significant coefficient.

2. We add another variable (in this case d80gnp) and the first variable becomes insignificant and the new variable is insignificant as well. Also R^2 hardly increases at all when the second variable is added.

3. This step isn't really necessary but just as insurance we do an F test of the hypothesis that the two co-linear variables have coefficients that are jointly insignificantly different from zero and we reject this hypothesis.

The nature of the co-linearity in this case is complex. It involves not only the presence of d80 and d80gnp but also the inclusion of the interest rate variable. For example if we eliminate the interest rate variable and run the regression we get:

```
regress cnsmpt gnp d80 d80gnp
```

Source	SS	df	MS	Number of obs	=	20
Model	11925011.8	3	3975003.94	F(3, 16)	=	9745.94
Residual	6525.80171	16	407.862607	Prob > F	=	0.0000
Total	11931537.6	19	627975.664	R-square	=	0.9995
				Adj R-square	=	0.9994
				Root MSE	=	20.196

cnsmp	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
gnp	.6315815	.0115979	54.457	0.000	.6069951 .6561678
d80	-203.2679	48.80912	-4.165	0.001	-306.7386 -99.79718
d80gnp	.0762695	.0169569	4.498	0.000	.0403225 .1122166
_cons	-5.479687	18.43501	-0.297	0.770	-44.56016 33.60079

Both coefficients for d80 and for d80gnp are now significant, but R^2 is lower. Further if we delete d80 rather than the interest rate we get:

```
. regress cnsmpt gnp intrate d80gnp
```

Source	SS	df	MS	Number of obs	=	20
Model	11927696.6	3	3975898.85	F(3, 16)	=	16561.64
Residual	3841.06748	16	240.066717	Prob > F	=	0.0000
Total	11931537.6	19	627975.664	R-square	=	0.9997
				Adj R-square	=	0.9996
				Root MSE	=	15.494

cnsmp	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
gnp	.6497492	.0083082	78.206	0.000	.6321367 .6673618
intrate	-7.334137	1.150334	-6.376	0.000	-9.772736 -4.895538
d80gnp	.018238	.0054272	3.360	0.004	.0067328 .0297432
_cons	25.27508	16.10374	1.570	0.136	-8.863326 59.41349

Now d80gnp coefficient is significant. So the multi-collinearity involves the three variables, intrate, d80 and d80gnp. Entering any two of the three yields significant coefficients for them but if all three are entered then we get multi-collinearity.