

interaction term

```
gen x3=x1*x2
```

```
reg y x1 x2 x3
```

multiple items

```
g x1=x^3
```

```
reg y x x1
```

heteroskedasticity

```
. clear all
```

```
. global root "C:\Users\Sabrina\Desktop\econometrics"
```

```
. cd "$root"
```

```
C:\Users\Sabrina\Desktop\econometrics
```

```
. use nerlove.dta
```

```
(Nerlove 1963 paper)
```

test

1.residual plot

```
. reg lntc lnq lnpl lnpl lnpl lnpl
```

Source	SS	df	MS	Number of obs	=	145
Model	269.524728	4	67.3811819	F(4, 140)	=	437.90
Residual	21.5420958	140	.153872113	Prob > F	=	0.0000
				R-squared	=	0.9260
				Adj R-squared	=	0.9239
Total	291.066823	144	2.02129738	Root MSE	=	.39227

lntc	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnq	.7209135	.0174337	41.35	0.000	.6864462	.7553808
lnpl	.4559645	.299802	1.52	0.131	-.1367602	1.048689
lnpk	-.2151476	.3398295	-0.63	0.528	-.8870089	.4567136
lnpf	.4258137	.1003218	4.24	0.000	.2274721	.6241554
_cons	-3.566513	1.779383	-2.00	0.047	-7.084448	-.0485779

```
. rvfplot
```

```

. graph save sca1,replace
(file sca1.gph saved)

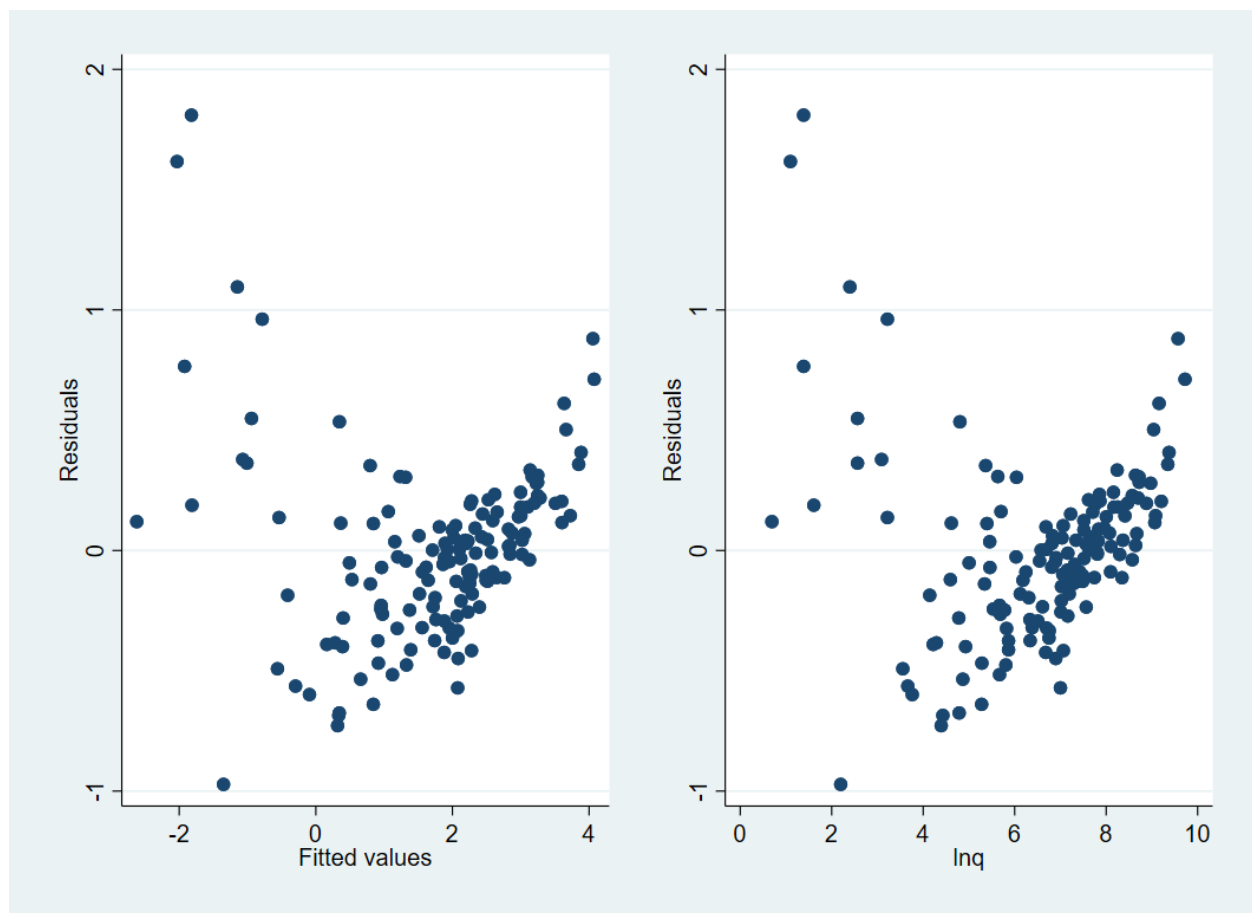
. rvppplot lnq

. graph save sca2,replace
(file sca2.gph saved)

. graph combine sca1.gph sca2.gph

. graph export sca3.png,replace
(file sca3.png written in PNG format)

```



scatter

2.White Test

```

. estat imtest,white

```

White's test for H_0 : homoskedasticity
 against H_a : unrestricted heteroskedasticity

chi2(14) = 73.88

Prob > chi2 = 0.0000

Cameron & Trivedi's decomposition of IM-test

Source	chi2	df	p
Heteroskedasticity	73.88	14	0.0000
Skewness	22.79	4	0.0001
Kurtosis	2.62	1	0.1055
Total	99.29	19	0.0000

```
. ssc install whitetst
checking whitetst consistency and verifying not already installed...
all files already exist and are up to date.
```

```
. whitetst
```

White's general test statistic : 73.8771 Chi-sq(14) P-value = 3.8e-10

3.BP Test

```
. estat hettest
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: fitted values of lntc

chi2(1) = 96.53

Prob > chi2 = 0.0000

```
. estat hettest,rhs
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: lnq lnpl lnpl lnpl lnpl

chi2(4) = 119.83

Prob > chi2 = 0.0000

```
. estat hettest lnq
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: lnq

chi2(1) = 106.35

Prob > chi2 = 0.0000

```
. estat hettest, iid
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: fitted values of lntc

chi2(1) = 29.13

Prob > chi2 = 0.0000

```
. estat hettest,rhs iid
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: lnq lnpl lnpg lnpr

chi2(4) = 36.16

Prob > chi2 = 0.0000

```
. estat hettest lnq, iid
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: lnq

chi2(1) = 32.10

Prob > chi2 = 0.0000

correction

OLS + robust

```
. reg lntc lnq lnpl lnpg lnpr,r
```

Linear regression

Number of obs = 145
F(4, 140) = 177.19
Prob > F = 0.0000
R-squared = 0.9260
Root MSE = .39227

lntc	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
lnq	.7209135	.0325376	22.16	0.000	.656585	.785242
lnpl	.4559645	.260326	1.75	0.082	-.0587139	.9706429
lnpg	-.2151476	.3233711	-0.67	0.507	-.8544698	.4241745
lnpr	.4258137	.0740741	5.75	0.000	.2793653	.5722622
_cons	-3.566513	1.718304	-2.08	0.040	-6.963693	-.1693331

WLS

```
. quietly reg lntc lnq lnpl lnpg lnpgf
. predict e11,res
. g e22=e11^2
. g lne22=log(e22)
. reg lne22 lnq,noc
```

Source	SS	df	MS	Number of obs	=	145
Model	2065.53636	1	2065.53636	F(1, 144)	=	419.95
Residual	708.275258	144	4.91857818	Prob > F	=	0.0000
				R-squared	=	0.7447
				Adj R-squared	=	0.7429
Total	2773.81162	145	19.1297353	Root MSE	=	2.2178

lne22	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnq	-.5527533	.0269733	-20.49	0.000	-.6060681	-.4994384

```
. predict lne22f
(option xb assumed; fitted values)
```

```
. g e22f = exp(lne22f)
```

```
. reg lntc lnq lnpl lnpg lnpgf [aw=1/e22f]
(sum of wgt is 8,181.07969295173)
```

Source	SS	df	MS	Number of obs	=	145
Model	173.069988	4	43.2674971	F(4, 140)	=	895.03
Residual	6.76790874	140	.048342205	Prob > F	=	0.0000
				R-squared	=	0.9624
				Adj R-squared	=	0.9613
Total	179.837897	144	1.24887428	Root MSE	=	.21987

lntc	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnq	.8759035	.0153841	56.94	0.000	.8454883	.9063187
lnpl	.5603879	.1734141	3.23	0.002	.2175389	.9032369
lnpg	-.0929807	.1960402	-0.47	0.636	-.4805627	.2946014
lnpgf	.4672438	.0616476	7.58	0.000	.3453632	.5891243
_cons	-5.522088	.9928472	-5.56	0.000	-7.485	-3.559176

Logit & Probit Model

command:

probit: `probit y x1 x2 x3,r`

logit: `logit y x1 x2 x3 , or vce(cluster clustvar)`

examples:

```
. use "$root\11.6\womenwk.dta",clear
```

```
. reg work age married children education,r
```

Linear regression	Number of obs	=	2,000
	F(4, 1995)	=	192.58
	Prob > F	=	0.0000
	R-squared	=	0.2026
	Root MSE	=	.41992

work	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
age	.0102552	.0012236	8.38	0.000	.0078556	.0126548
married	.1111116	.0226719	4.90	0.000	.0666485	.1555748
children	.1153084	.0056978	20.24	0.000	.1041342	.1264827
education	.0186011	.0033006	5.64	0.000	.0121282	.025074
_cons	-.2073227	.0534581	-3.88	0.000	-.3121622	-.1024832

```
. logit work age married children education,nolog
```

Logistic regression	Number of obs	=	2,000
	LR chi2(4)	=	476.62
	Prob > chi2	=	0.0000
Log likelihood = -1027.9144	Pseudo R2	=	0.1882

work	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
age	.0579303	.007221	8.02	0.000	.0437773	.0720833
married	.7417775	.1264705	5.87	0.000	.4938998	.9896552
children	.7644882	.0515289	14.84	0.000	.6634935	.865483
education	.0982513	.0186522	5.27	0.000	.0616936	.134809
_cons	-4.159247	.3320401	-12.53	0.000	-4.810034	-3.508461

```
. logit work age married children education, r nolog
```

Logistic regression	Number of obs	=	2,000
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Log pseudolikelihood = -1027.9144

Wald chi2(4) = 344.54
 Prob > chi2 = 0.0000
 Pseudo R2 = 0.1882

work	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
age	.0579303	.0072054	8.04	0.000	.0438079	.0720527
married	.7417775	.1272191	5.83	0.000	.4924326	.9911224
children	.7644882	.0497584	15.36	0.000	.6669635	.8620129
education	.0982513	.019011	5.17	0.000	.0609904	.1355121
_cons	-4.159247	.327398	-12.70	0.000	-4.800936	-3.517559

. logit work age married children education, or nolog

Logistic regression

Number of obs = 2,000
 LR chi2(4) = 476.62
 Prob > chi2 = 0.0000
 Pseudo R2 = 0.1882

Log likelihood = -1027.9144

work	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
age	1.059641	.0076517	8.02	0.000	1.04475	1.074745
married	2.099664	.2655457	5.87	0.000	1.638694	2.690307
children	2.147895	.1106786	14.84	0.000	1.941563	2.376153
education	1.10324	.0205779	5.27	0.000	1.063636	1.144318
_cons	.0156193	.0051862	-12.53	0.000	.0081476	.029943

Note: _cons estimates baseline odds.

. margins, dydx(*)

Average marginal effects

Model VCE : OIM

Number of obs = 2,000

Expression : Pr(work), predict()
 dy/dx w.r.t. : age married children education

	Delta-method dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]	
age	.0099674	.0011682	8.53	0.000	.0076778	.0122569
married	.127629	.021152	6.03	0.000	.0861717	.1690862
children	.1315365	.007073	18.60	0.000	.1176736	.1453994
education	.0169049	.0031243	5.41	0.000	.0107814	.0230285

```
. margins,dydx(*) atmeans
```

```
Conditional marginal effects      Number of obs      =      2,000
Model VCE      : OIM
```

```
Expression      : Pr(work), predict()
dy/dx w.r.t.    : age married children education
at              : age          =      36.208 (mean)
                  married      =      .6705 (mean)
                  children      =      1.6445 (mean)
                  education     =      13.084 (mean)
```

	Delta-method					
	dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]	
age	.0115031	.0014236	8.08	0.000	.0087129	.0142934
married	.1472934	.0248209	5.93	0.000	.0986453	.1959415
children	.151803	.0093768	16.19	0.000	.1334249	.1701812
education	.0195096	.0036991	5.27	0.000	.0122596	.0267596

```
. margins,dydx(*) at(age==30)
```

```
Average marginal effects      Number of obs      =      2,000
Model VCE      : OIM
```

```
Expression      : Pr(work), predict()
dy/dx w.r.t.    : age married children education
at              : age          =      30
```

	Delta-method					
	dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]	
age	.011179	.0014719	7.59	0.000	.008294	.0140639
married	.1431427	.0232525	6.16	0.000	.0975687	.1887167
children	.1475253	.0074033	19.93	0.000	.1330151	.1620355
education	.0189598	.0034727	5.46	0.000	.0121534	.0257662

```
. logit work age married children education , nolog vce(cluster age)
```

```
Logistic regression      Number of obs      =      2,000
                          Wald chi2(4)      =      576.81
                          Prob > chi2       =      0.0000
Log pseudolikelihood = -1027.9144      Pseudo R2      =      0.1882
```

(Std. Err. adjusted for 40 clusters in age)

	Robust
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