

# Investigating the effects of factor variables

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This workshop's materials have been posted to the following website for your convenience.

<http://www.stata.com/users/jpitblado/2011germany/>

## 1 What are effects?

## Effects

The effect of a factor variable is the change in a measurement between two or more levels of the factor.

Example:

- Difference in average cholesterol measurement between two age groups in a population.

## 2 Computing effects

```
. webuse cholesterol
(Artificial cholesterol data)
```

```
. describe chol agegrp
```

variable name	storage type	display format	value label	variable label
chol	float	%9.0g		cholesterol level (mg/dL)
agegrp	float	%9.0g	ages	
. label list ages				
ages:				
	1	10-19		
	2	20-29		
	3	30-39		
	4	40-59		
	5	60-79		

```
. regress chol i.agegrp
```

Source	SS	df	MS
Model	14943.3997	4	3735.84993
Residual	7468.21971	70	106.688853
Total	22411.6194	74	302.859722

Number of obs = 75  
F( 4, 70) = 35.02  
Prob > F = 0.0000  
R-squared = 0.6668  
Adj R-squared = 0.6477  
Root MSE = 10.329

chol	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
agegrp						
2	8.203575	3.771628	2.18	0.033	.6812991	15.72585
3	21.54105	3.771628	5.71	0.000	14.01878	29.06333
4	30.15067	3.771628	7.99	0.000	22.6284	37.67295
5	38.76221	3.771628	10.28	0.000	31.23993	46.28448
_cons	180.5198	2.666944	67.69	0.000	175.2007	185.8388

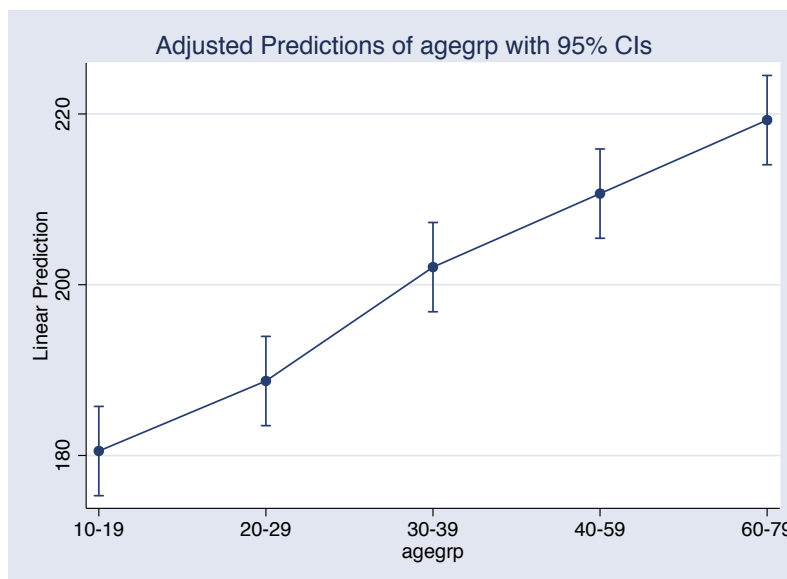
```
. margins agegrp
```

Adjusted predictions	Number of obs	=	75
Model VCE : OLS			
Expression : Linear prediction, predict()			

Delta-method

	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
agegrp						
1	180.5198	2.666944	67.69	0.000	175.2926	185.7469
2	188.7233	2.666944	70.76	0.000	183.4962	193.9504
3	202.0608	2.666944	75.76	0.000	196.8337	207.2879
4	210.6704	2.666944	78.99	0.000	205.4433	215.8975
5	219.282	2.666944	82.22	0.000	214.0548	224.5091

```
. marginsplot
```



## Coefficient table

**regress** reports some simple tests on the effects of **agegrp** on **chol**.

## How can we change the base level?

- Refit the model using the **b.** operator.
- Use **test** or **lincom** to perform the comparison.

```
. lincom 1.agegrp - 5.agegrp
( 1) 1b.agegrp - 5.agegrp = 0
```

chol	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
(1)	-38.76221	3.771628	-10.28	0.000	-46.28448	-31.23993

```
. lincom 2.agegrp - 5.agegrp
( 1) 2.agegrp - 5.agegrp = 0
```

chol	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
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(1)	-30.55863	3.771628	-8.10	0.000	-38.08091	-23.03636
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- New in Stata 12
- ANOVA-style tests of linear hypotheses involving factor variables and their interactions from the most recently fit model.
  - main effects
  - simple effects
  - interaction effects
  - nested effects
- Decompose tests into individual components/effects/constrasts.
  - built-in contrast operators
  - user defined contrasts

## Syntax

**contrast** *op.varname* [, *options*]

<i>op.</i>	Description
<b>r.</b>	diff from a reference (base) level; the default
<b>a.</b>	diff from the next level (adjacent)
<b>ar.</b>	diff from the previous level (reverse adjacent)

```
. contrast rb5.agegrp, effects
Contrasts of marginal linear predictions
Margins      : asbalanced
```

	df	F	P>F
agegrp			
(1 vs 5)	1	105.62	0.0000
(2 vs 5)	1	65.65	0.0000
(3 vs 5)	1	20.85	0.0000
(4 vs 5)	1	5.21	0.0255
Joint	4	35.02	0.0000
Residual	70		

	Contrast	Std. Err.	t	P> t	[95% Conf. Interval]	
agegrp						
(1 vs 5)	-38.76221	3.771628	-10.28	0.000	-46.28448	-31.23993
(2 vs 5)	-30.55863	3.771628	-8.10	0.000	-38.08091	-23.03636
(3 vs 5)	-17.22115	3.771628	-4.57	0.000	-24.74343	-9.698877
(4 vs 5)	-8.611533	3.771628	-2.28	0.025	-16.13381	-1.089257

```
. contrast a.agegrp, effects
Contrasts of marginal linear predictions
Margins      : asbalanced
```

	df	F	P>F
agegrp			
(1 vs 2)	1	4.73	0.0330
(2 vs 3)	1	12.51	0.0007
(3 vs 4)	1	5.21	0.0255
(4 vs 5)	1	5.21	0.0255
Joint	4	35.02	0.0000
Residual	70		

	Contrast	Std. Err.	t	P> t	[95% Conf. Interval]	
agegrp						
(1 vs 2)	-8.203575	3.771628	-2.18	0.033	-15.72585	-.6812991
(2 vs 3)	-13.33748	3.771628	-3.54	0.001	-20.85976	-5.815204
(3 vs 4)	-8.60962	3.771628	-2.28	0.025	-16.1319	-1.087345
(4 vs 5)	-8.611533	3.771628	-2.28	0.025	-16.13381	-1.089257

## As-balanced effects

Compute effects using linear combinations that weight each margin equally.

### *op.* Description

- 
- g.** diff from the balanced grand mean
  - h.** diff from the balanced mean of subsequent levels (Helmert)
  - j.** diff from the balanced mean of previous levels (reverse Helmert)
  - p.** orthogonal polynomial in the level values
  - q.** orthogonal polynomial in the level sequence
- 

```
. contrast h.agegrp, effects
Contrasts of marginal linear predictions
Margins      : asbalanced
```

	df	F	P>F
agegrp			
(1 vs >1)	1	68.42	0.0000
(2 vs >2)	1	50.79	0.0000
(3 vs >3)	1	15.63	0.0002
(4 vs 5)	1	5.21	0.0255
Joint	4	35.02	0.0000
Residual	70		

	Contrast	Std. Err.	t	P> t	[95% Conf. Interval]	
agegrp						
(1 vs >1)	-24.66438	2.981734	-8.27	0.000	-30.61126	-18.7175
(2 vs >2)	-21.94774	3.079522	-7.13	0.000	-28.08965	-15.80583
(3 vs >3)	-12.91539	3.266326	-3.95	0.000	-19.42987	-6.400905
(4 vs 5)	-8.611533	3.771628	-2.28	0.025	-16.13381	-1.089257

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### As-observed effects

Compute effects using linear combinations that weight each margin according to the sample frequencies of the levels.

<i>op.</i>	Description
<b>gw.</b>	diff from the weighted grand mean
<b>hw.</b>	diff from the weighted mean of subsequent levels (Helmert)
<b>jw.</b>	diff from the weighted mean of previous levels (reverse Helmert)
<b>pw.</b>	weighted orthogonal polynomial in the level values
<b>qw.</b>	weighted orthogonal polynomial in the level squence

## 3 Higher order effects

### Interaction effect

When the effect of one factor depends on the level of other factors.

Example:

- Difference in average blood pressure measurement between two dosage levels for men and women.
- Factors: dosage and gender

```
. webuse bpchange
(Artificial blood pressure data)
. describe
Contains data from http://localpress.stata.com/data/r12/bpchange.dta
  obs:      30      Artificial blood pressure data
  vars:      3      21 Feb 2011 16:59
  size:     360
```

variable name	storage type	display format	value label	variable label
bpchange	float	%9.0g		change in diastolic blood pressure
dose	float	%9.0g		dosage in milligrams per day
gender	float	%9.0g	gender	

Sorted by:

```
. anova bpchange dose##gender
```

	Number of obs =	30	R-squared =	0.9647	
	Root MSE =	1.4677	Adj R-squared =	0.9573	
Source	Partial SS	df	MS	F	Prob > F
Model	1411.9087	5	282.381741	131.09	0.0000

dose	963.481795	2	481.740897	223.64	0.0000
gender	355.118817	1	355.118817	164.85	0.0000
dose#gender	93.3080926	2	46.6540463	21.66	0.0000
Residual	51.699253	24	2.15413554		
Total	1463.60796	29	50.4692399		

```
. contrast dose#gender
Contrasts of marginal linear predictions
Margins      : asbalanced
```

	df	F	P>F
dose#gender	2	21.66	0.0000
Residual	24		

```
. contrast dose##gender
Contrasts of marginal linear predictions
Margins      : asbalanced
```

	df	F	P>F
dose	2	223.64	0.0000
gender	1	164.85	0.0000
dose#gender	2	21.66	0.0000
Residual	24		

```
. contrast a.dose@gender, effects
Contrasts of marginal linear predictions
Margins      : asbalanced
```

	df	F	P>F
dose@gender			
(250 vs 500) 1	1	47.24	0.0000
(250 vs 500) 2	1	122.90	0.0000
(500 vs 750) 1	1	11.06	0.0028
(500 vs 750) 2	1	70.68	0.0000
Joint	4	122.65	0.0000
Residual	24		

	Contrast	Std. Err.	t	P> t	[95% Conf. Interval]	
dose@gender						
(250 vs 500) 1	6.380018	.9282533	6.87	0.000	4.464198	8.295839
(250 vs 500) 2	10.29073	.9282533	11.09	0.000	8.374914	12.20655
(500 vs 750) 1	3.087217	.9282533	3.33	0.003	1.171396	5.003038
(500 vs 750) 2	7.803784	.9282533	8.41	0.000	5.887963	9.719605

- Factor effects on slopes
  - `contrast fvar#c.xvar`
- Nonlinear models
  - `clogit, glm, logit, heckman, ivregress, nbreg, poisson, ...`
- Multiple equations
  - `manova, mlogit, mprobit, mvreg`
  - Special `_eqns` factor for effects between equations
- Adjusting for multiple comparisons
  - Bonferroni
  - Šidák
- [R] **contrast** — over 50 pages of informaton

## 4 Pairwise comparisons

### Syntax

`pwcompare marginlist [ , options ]`

- Intercept and slope effects
- Nonlinear models
- Multiple equations
- Adjusting for multiple comparisons
  - Generally applicable
    - \* Bonferroni, Scheffe, Šidák
  - Linear models only
    - \* Tukey, Student-Newman-Keuls, Duncan, Dunnett
- [R] **pwcompare** — almost 30 pages of informaton



## 5 Summary

- **marginsplot** graphs results from **margins**
- **contrast** provides a short and simple syntax for testing all kinds of factor effects
- **pwcompare** performs pairwise comparisons of marginal linear predictions
- **margins** has new **contrast** and **pwcompare** features