

How can I use the margins command in Stata to obtain simple main effects in an ANOVA?

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The margins command in Stata is a useful tool for obtaining simple main effects in an ANOVA (Analysis of Variance) analysis. It allows for the calculation of the average predicted values for each level of a categorical variable, holding all other variables at their means. This command is particularly helpful in ANOVA models with multiple factors, as it can provide a clearer understanding of the effects of each individual factor on the outcome variable. By using the margins command, researchers can easily examine the simple main effects of each factor and make informed decisions about their data analysis.

How can I get anova simple main effects with the margins command? Stata FAQ

Now that Stata 11 has been released, you be wondering if there is an easy way to compute tests of simple main effects? Yes there is and this FAQ will show you how to get tests of simple main effects using the new margins command.

Consider a two-factor design using a dataset called crf24.

**use <https://stats.idre.ucla.edu/stat/stata/faq/crf24>, clear
anova y a##b**

Number of obs = 32 R-squared = 0.9214

Root MSE = .877971 Adj R-squared = 0.8985

Source | Partial SS df MS F Prob > F

-----+-----

Model | 217 7 31 40.22 0.0000

|

a | 3.125 1 3.125 4.05 0.0554

b | 194.5 3 64.83333333 84.11 0.0000

a#b | 19.375 3 6.458333333 8.38 0.0006

|

Residual | 18.5 24 .7708333333

-----+-----

Total | 235.5 31 7.59677419

As you can see the a#b interaction is statistically significant.

Next we will use the margins command on the a#b interaction. We will include the post option and if the design were unbalanced or included a covariate we would also

include the asbalanced option.

`margins a#b, post`

Adjusted predictions Number of obs = 32

Expression : Linear prediction, predict()

| Delta-method

| Margin Std. Err. z P>|z|

a#b |

1 1 | 3.75 .4389856 8.54 0.000 2.889604 4.610396

1 2 | 4 .4389856 9.11 0.000 3.139604 4.860396

1 3 | 7 .4389856 15.95 0.000 6.139604 7.860396

1 4 | 8 .4389856 18.22 0.000 7.139604 8.860396

2 1 | 1.75 .4389856 3.99 0.000 .8896041 2.610396

2 2 | 3 .4389856 6.83 0.000 2.139604 3.860396

2 3 | 5.5 .4389856 12.53 0.000 4.639604 6.360396

2 4 | 10 .4389856 22.78 0.000 9.139604 10.8604

The margins command has given us a list of the cell means and their standard errors.

We can use combinations of these means to compute the test of simple main effects. For example, a1 versus a2 at b1 would test cell a1,b1 versus cell a2,b1. Here is what the tests of simple main effects for a each level of b would look like.

test 1.a#1.b == 2.a#1.b

(1) 1bn.a#1bn.b - 2.a#1bn.b = 0

chi2(1) = 10.38

Prob > chi2 = 0.0013

test 1.a#2.b == 2.a#2.b

(1) 1bn.a#2.b - 2.a#2.b = 0

chi2(1) = 2.59

Prob > chi2 = 0.1072

test 1.a#3.b == 2.a#3.b

(1) 1bn.a#3.b - 2.a#3.b = 0

chi2(1) = 5.84

Prob > chi2 = 0.0157

test 1.a#4.b == 2.a#4.b

(1) 1bn.a#4.b - 2.a#4.b = 0

chi2(1) = 10.38

Prob > chi2 = 0.0013

Each one of the above tests used only one degree of freedom so the terms in the test were simple. We will next move on to test of b at each level of a. Each of these tests have three degrees of freedom and so, will will

involve a more complex test command.

We will start off showing the accum approach and then show a cleaner way to get the result with a single test command.

```
/* test of b at a==1 */
```

```
test 1.a#1.b == 1.a#2.b
```

```
test 1.a#1.b == 1.a#3.b, accum
```

```
test 1.a#1.b == 1.a#4.b, accum
```

```
( 1) 1bn.a#1bn.b - 1bn.a#2.b = 0
```

```
( 2) 1bn.a#1bn.b - 1bn.a#3.b = 0
```

```
( 3) 1bn.a#1bn.b - 1bn.a#4.b = 0
```

```
chi2( 3) = 70.95
```

```
Prob > chi2 = 0.0000
```

```
/* the same test using a single test command */
```

```
test (1.a#1.b == 1.a#2.b)(1.a#1.b == 1.a#3.b)(1.a#1.b == 1.a#4.b)
```

(1) 1bn.a#1bn.b - 1bn.a#2.b = 0

(2) 1bn.a#1bn.b - 1bn.a#3.b = 0

(3) 1bn.a#1bn.b - 1bn.a#4.b = 0

chi2(3) = 70.95

Prob > chi2 = 0.0000

/* convert chi2 to approximate F-value */

display r(chi2)/r(df)

23.648649

We will finish up with tests of b at a==2 just by changing all of the 1.a terms to 2.a

/* test of b at a==1 */

test (2.a#1.b == 2.a#2.b)(2.a#1.b == 2.a#3.b)(2.a#1.b == 2.a#4.b)

(1) 2.a#1bn.b - 2.a#2.b = 0

(2) 2.a#1bn.b - 2.a#3.b = 0

(3) 2.a#1bn.b - 2.a#4.b = 0

chi2(3) = 206.51

Prob > chi2 = 0.0000

/* convert chi2 to approximate F-value */

display r(chi2)/r(df)

68.837838

This FAQ only covers the computation of the tests of simple main effects using the margins command. The FAQ does not cover computing the critical values of these tests. There is a user written ado-program smecriticalvalue which can assist in this process (search smecriticalvalue).

Let's try a little more challenging example, one that is unbalanced with a covariate.

use <https://stats.idre.ucla.edu/stat/data/hsbdemo>, clear

anova write female##prog c.read

Number of obs = 200 R-squared = 0.4889

Root MSE = 6.88105 Adj R-squared = 0.4730

Source | Partial SS df MS F Prob > F

Model	 	8740.54801	6	1456.758	30.77	0.0000
 						
female	 	1680.08743	1	1680.08743	35.48	0.0000
prog	 	632.133455	2	316.066727	6.68	0.0016
female#prog	 	301.770797	2	150.885399	3.19	0.0435
read	 	4110.18709	1	4110.18709	86.81	0.0000
 						
Residual	 	9138.32699	193	47.3488445		

Total | 17878.875 199 89.843593

The female#prog interaction is significant but we'll go ahead and compute the tests of simple main effects. This time because the model is both unbalanced and includes a covariate we will include the asbalanced option which is equivalent to least squares means (lsmeans) or expected marginal means (emmeans) in SAS and SPSS.

margins female#prog, asbalanced post

Predictive margins Number of obs = 200

Expression : Linear prediction, predict()

| Delta-method

| Margin Std. Err. z P>|z|

-----+-----
female#prog |

0 1 | 48.7854 1.502058 32.48 0.000 45.84142 51.72938

0 2 | 52.61465 1.026455 51.26 0.000 50.60284 54.62647

0 3 | 45.08098 1.476716 30.53 0.000 42.18667 47.97529

1 1 | 55.85857 1.432221 39.00 0.000 53.05146 58.66567

1 2 | 55.68658 .926245 60.12 0.000 53.87117 57.50199

1 3 | 53.71585 1.356821 39.59 0.000 51.05653 56.37517

/* tests of female at each level of prog */

test 0.female#1.prog == 1.female#1.prog

(1) 0bn.female#1bn.prog - 1.female#1bn.prog = 0

chi2(1) = 11.56

Prob > chi2 = 0.0007

test 0.female#2.prog == 1.female#2.prog

(1) 0bn.female#2.prog - 1.female#2.prog = 0

chi2(1) = 5.17

Prob > chi2 = 0.0229

test 0.female#3.prog == 1.female#3.prog

(1) 0bn.female#3.prog - 1.female#3.prog = 0

chi2(1) = 19.54

Prob > chi2 = 0.0000

/* tests of prog at each level of female */

test (0.female#1.prog == 0.female#2.prog)(0.female#1.prog == 0.female#3.prog)

(1) 0bn.female#1bn.prog - 0bn.female#2.prog = 0

(2) 0bn.female#1bn.prog - 0bn.female#3.prog = 0

chi2(2) = 17.41

Prob > chi2 = 0.0002

/* convert chi2 to approximate F-value */

display r(chi2)/r(df)

8.7026677

```
test          (1.female#1.prog          ==  
1.female#2.prog)(1.female#1.prog == 1.female#3.prog)
```

```
( 1) 1.female#1bn.prog - 1.female#2.prog = 0
```

```
( 2) 1.female#1bn.prog - 1.female#3.prog = 0
```

```
chi2( 2) = 1.69
```

```
Prob > chi2 = 0.4287
```

```
/* convert chi2 to approximate F-value */
```

```
display r(chi2)/r(df)
```

```
.84693839
```

So that's how to compute tests of simple main effects using margins in Stata 11.