STAT 401A - Statistical Methods for Research Workers Two-way ANOVA

Jarad Niemi (Dr. J)

Iowa State University

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Data

An experiment was run on tomato plants to determine the effect of

- 3 different varieties (A,B,C) and
- 4 different planting densities (10,20,30,40)

on yield.

There is an expectation that planting density will have a different effect depending on the variety. Therefore a balanced, complete, randomized design was used.

- complete: each treatment (variety × density) is represented in the experiment
- balanced: each treatment in the experiment has the same number of replications
- randomized: treatment was randomly assigned to the plot

This is also referred to as a full factorial or fully crossed design.

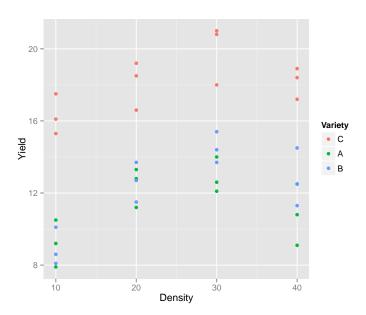
Hypotheses

- Does variety affect mean yield?
 - Is the mean yield for variety A different from B on average?
 - Is the mean yield for variety A different from B at a particular value for density?
- Does density affect mean yield?
 - Is the mean yield for density 10 different from density 20 on average?
 - Is the mean yield for density 10 different from density 20 at a particular value for variety?
- Does density affect yield differently for each variety?

For all of these questions, we want to know

- is there any effect and
- if yes, what is the nature of the effect.

Confidence intervals can answer these questions.



Summary statistics

Number of replicates

Mean Yield

```
Variety 10 20 30 40
1 C 16.300000 18.10000 19.93333 18.16667
2 A 9.2000000 12.43333 12.90000 10.80000
3 B 8.933333 12.63333 14.50000 12.76667
```

Standard deviation of yield

```
Variety 10 20 30 40

1 C 1.113553 1.345362 1.6772994 0.8736895

2 A 1.300000 1.096966 0.9848858 1.7000000

3 B 1.040833 1.101514 0.8544004 1.6165808
```

Two-way ANOVA

- ullet Setup: Two categorical explanatory variables with I and J levels
- Model:

$$Y_{ijk} \stackrel{ind}{\sim} N(\mu_{ij}, \sigma^2)$$

where Y_{ijk} is the

- kth observation at the
- ith level of variable 1 (variety) with i = 1, ..., I and the
- jth level of variable 2 (density) with j = 1, ..., J.

Consider the models:

- Additive: $\mu_{ij} = \mu + \nu_i + \delta_i$
- Cell-means: $\mu_{ij} = \mu + \nu_i + \delta_j + \gamma_{ij}$

	10	20	30	40
Α	μ_{11}	μ_{12}	μ_{13}	μ_{14}
В	μ_{21}	μ_{22}	μ_{23}	μ_{24}
С	μ_{31}	μ_{32}	μ_{33}	μ_{34}

As a regression model

- Assign a reference level for both variety (C) and density (40).
- ② Let V_i and D_i be the variety and density for observation i.
- **3** Build indicator variables, e.g. $I(V_i = A)$ and $I(D_i = 10)$.
- The additive model:

$$\mu_i = \beta_0 + \beta_1 I(V_i = A) + \beta_2 I(V_i = B) + \beta_3 I(D_i = 10) + \beta_4 I(D_i = 20) + \beta_5 I(D_i = 30).$$

 β_1 is the expected difference in yield between varieties A and C at any fixed density

The cell-means model:

$$\mu_{i} = \beta_{0} + \beta_{1}I(V_{i} = A) + \beta_{2}I(V_{i} = B) + \beta_{3}I(D_{i} = 10) + \beta_{4}I(D_{i} = 20) + \beta_{5}I(D_{i} = 30)$$

$$+\beta_{6}I(V_{i} = A)I(D_{i} = 10) + \beta_{7}I(V_{i} = A)I(D_{i} = 20) + \beta_{8}I(V_{i} = A)I(D_{i} = 30) + \beta_{9}I(V_{i} = B)I(D_{i} = 10) + \beta_{10}I(V_{i} = B)I(D_{i} = 20) + \beta_{11}I(V_{i} = B)I(D_{i} = 30)$$

 β_1 is the expected difference in yield between varieties A and C at a density of 40

ANOVA Table

ANOVA Table - Additive model

Source	SS	df	MS	F
Factor A	SSA	(I-1)	SSA/(I-1)	MSA/MSE
Factor B	SSB	(J-1)	SSB/(J-1)	MSB/MSE
Error	SSE	n-I-J-1	SSE/(n-I-J-1)	
Total	SST	n-1		

ANOVA Table - Cell-means model

Source	SS	df	MS	
Factor A	SSA	I-1	SSA/(I-1)	MSA/MSE
Factor B	SSB	J-1	$SSB/(\mathrm{J}\text{-}1)$	MSB/MSE
Interaction AB	SSAB	(I-1)(J-1)	SSAB /(I-1)(J-1)	MSAB/MSE
Error	SSE	n-IJ	$SSE/(n\text{-}\mathrm{IJ})$	
Total	SST	n-1		

Additive vs cell-means

Opinions differ on whether to use an additive vs a cell-means model when the interaction is not significant. Remember that an insignificant test does not prove that there is no interaction.

	Additive	Cell-means
Interpretation	Direct	Complicated
Estimate of σ^2	Biased	Unbiased

We will continue using the cell-means model to answer the scientific questions of interest.

```
DATA tomato;
  INFILE 'Ch13-tomato.csv' DSD FIRSTOBS=2;
  INPUT variety $ density yield;
PROC GLM DATA=tomato PLOTS=all:
  CLASS variety density;
  MODEL yield = variety|density / SOLUTION;
  LSMEANS variety / cl adjust=tukey;
  LSMEANS density / cl adjust=tukey;
  LSMEANS variety*density / cl adjust=tukey;
  RUN;
```

The GLM Procedure

Dependent Variable: yield

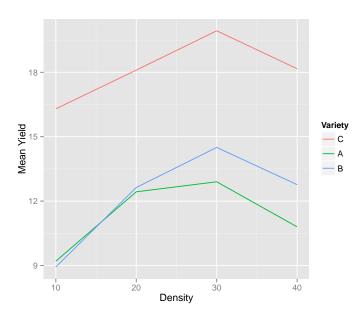
			Sun	ı of					
Source		DF	Squa	res	Mean	Square	F	Value	Pr > F
Model		11	422.3155	5556	38.3	3923232		24.22	<.0001
Error		24	38.0400	0000	1.5	5850000			
Corrected Total		35	460.3555	5556					
	R-Square	Coeff	Var	Root M	ISE	yield Me	an		
	0.917368	9.06	4568	1.2589	968	13.888	389		
Source		DF	Type 1	SS	Mean	Square	F	Value	Pr > F
variety		2	327.5972	2222	163.7	7986111		103.34	<.0001
density		3	86.6866	6667	28.8	3955556		18.23	<.0001
variety*density		6	8.0316	6667	1.3	3386111		0.84	0.5484
Source		DF	Type III	SS	Mean	Square	F	Value	Pr > F
variety		2	327.5972	2222	163.7	7986111	:	103.34	<.0001
density		3	86.6866	6667	28.8	3955556		18.23	<.0001
variety*density		6	8.0316	6667	1.3	3386111		0.84	0.5484

The Type I and Type III SS are equal because the design is balanced.

MODEL yield = variety|density / SOLUTION;

The GLM Procedure

				Standard		
Parameter		Estimate		Error	t Value	Pr > t
Intercept		18.16666667	В	0.72686542	24.99	<.0001
variety I	A	-7.36666667	В	1.02794293	-7.17	<.0001
variety H	В	-5.40000000	В	1.02794293	-5.25	<.0001
variety (C	0.00000000	В			
density 1	10	-1.8666667	В	1.02794293	-1.82	0.0819
density 2	20	-0.06666667	В	1.02794293	-0.06	0.9488
density 3	30	1.76666667	В	1.02794293	1.72	0.0986
density 4	40	0.00000000	В			
variety*density A	A 10	0.26666667	В	1.45373083	0.18	0.8560
variety*density A	A 20	1.70000000	В	1.45373083	1.17	0.2537
variety*density A	A 30	0.33333333	В	1.45373083	0.23	0.8206
variety*density A	A 40	0.00000000	В			
variety*density B	B 10	-1.96666667	В	1.45373083	-1.35	0.1887
variety*density B	B 20	-0.06666667	В	1.45373083	-0.05	0.9638
variety*density B	B 30	-0.03333333	В	1.45373083	-0.02	0.9819
variety*density B	B 40	0.00000000	В			
variety*density (C 10	0.00000000	В			



Is the mean yield for variety A different from B on average?

```
LSMEANS variety / cl adjust=tukey;
                                      Least Squares Means
                          Adjustment for Multiple Comparisons: Tukey
                            Least Squares Means for effect variety
                             Pr > |t| for HO: LSMean(i)=LSMean(i)
                                  Dependent Variable: yield
                        i/j
                                                 0.2249
                                                               < .0001
                                    0.2249
                                                               <.0001
                                    <.0001
                                                 <.0001
                     varietv
                                vield LSMEAN
                                                 95% Confidence Limits
                                   11.333333
                                                 10.583245
                                                               12.083422
                     Α
                                   12.208333
                                                  11.458245 12.958422
                                   18.125000
                                                  17.374912 18.875088
                             Least Squares Means for Effect variety
                                  Difference
                                                    Simultaneous 95%
                                                 Confidence Limits for
                                     Retween
                                                 LSMean(i)-LSMean(j)
                                       Means
                                  -0.875000
                                                  -2.158534 0.408534
                                  -6.791667
                                                  -8.075201 -5.508132
                                   -5.916667
                                                  -7.200201 -4.633132
```

Is the mean yield at density 10 different from density 20 on average?

```
LSMEANS density / cl adjust=tukey;
```

Least Squares Means
Adjustment for Multiple Comparisons: Tukey

density yield LSMEAN 95% Confidence Limits 11.477778 10.611650 12.343905 10 20 14.388889 13.522762 15.255016 15.777778 14.911650 16.643905 13.911111 13.044984 14.777238 40

Least Squares Means for Effect density Difference Simultaneous 95% Confidence Limits for Between i LSMean(i)-LSMean(i) Means -2.911111 -4.548299 -1.273923 -4.300000 -5.937188 -2.662812 -2.433333 -4.070521 -0.796145 -3.026077 0.248299 -1.388889 0.477778 -1.159410 2.114966

1.866667

0.229479

3.503855

Is mean yield different for particular combinations?

LSMEANS variety*density / cl adjust=tukey;

variety	density	yield LSMEAN	95% Confiden	ce Limits
A	10	9.200000	7.699824	10.700176
A	20	12.433333	10.933157	13.933510
A	30	12.900000	11.399824	14.400176
A	40	10.800000	9.299824	12.300176
В	10	8.933333	7.433157	10.433510
В	20	12.633333	11.133157	14.133510
В	30	14.500000	12.999824	16.000176
В	40	12.766667	11.266490	14.266843
C	10	16.300000	14.799824	17.800176
C	20	18.100000	16.599824	19.600176
C	30	19.933333	18.433157	21.433510
C	40	18.166667	16.666490	19.666843

Is mean yield different for particular combinations?

LSMEANS variety*density / cl adjust=tukey;

Least Squares Means for Effect variety*density

		Difference	Simultane	
		Between	Confidence L	
i	j	Means	LSMean(i)-L	SMean(j)
1	2	-3.233333	-6.939704	0.473037
1	3	-3.700000	-7.406371	0.006371
1	4	-1.600000	-5.306371	2.106371
1	5	0.266667	-3.439704	3.973037
1	6	-3.433333	-7.139704	0.273037
1	7	-5.300000	-9.006371	-1.593629
1	8	-3.566667	-7.273037	0.139704
1	9	-7.100000	-10.806371	-3.393629
1	10	-8.900000	-12.606371	-5.193629
1	11	-10.733333	-14.439704	-7.026963
1	12	-8.966667	-12.673037	-5.260296
2	3	-0.466667	-4.173037	3.239704
2	4	1.633333	-2.073037	5.339704
2	5	3.500000	-0.206371	7.206371
2	6	-0.200000	-3.906371	3.506371
2	7	-2.066667	-5.773037	1.639704
2	8	-0.333333	-4.039704	3.373037
2	9	-3.866667	-7.573037	-0.160296
2	10	-5.666667	-9.373037	-1.960296
2	11	-7.500000	-11.206371	-3.793629
2	12	-5.733333	-9.439704	-2.026963
3	4	2.100000	-1.606371	5.806371
3	5	3.966667	0.260296	7.673037
3	6	0.266667	-3.439704	3.973037

```
tomato$Density = factor(tomato$Density)
m = lm(Yield~Variety*Density, tomato)
anova(m)
Analysis of Variance Table
Response: Yield
              Df Sum Sq Mean Sq F value Pr(>F)
Variety
                2 327.60 163.799 103.3430 1.608e-12 ***
Density
               3 86.69 28.896 18.2306 2.212e-06 ***
Variety: Density 6 8.03 1.339 0.8445 0.5484
Residuals
         24 38.04 1.585
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
lsmeans(m, pairwise~Variety)
$1smeans
Variety 1smean SE df lower.CL upper.CL
C 18.12500 0.3634327 24 17.37491 18.87509
        11 33333 0 3634327 24 10 58325 12 08342
        12.20833 0.3634327 24 11.45825 12.95842
Results are averaged over the levels of: Density
Confidence level used: 0.95
$contrasts
contrast estimate
                         SE df t.ratio p.value
C - A 6.791667 0.5139715 24 13.214 <.0001
C - B 5.916667 0.5139715 24 11.512 <.0001
A - B -0.875000 0.5139715 24 -1.702 0.2249
Results are averaged over the levels of: Density
P value adjustment: tukey method for a family of 3 means
```

library(lsmeans)

```
lsmeans(m, pairwise Density)
$1smeans
Density 1smean
                     SE df lower.CL upper.CL
       11.47778 0.4196559 24 10.61165 12.34391
10
20
       14.38889 0.4196559 24 13.52276 15.25502
30
       15.77778 0.4196559 24 14.91165 16.64391
40
       13.91111 0.4196559 24 13.04498 14.77724
Results are averaged over the levels of: Variety
Confidence level used: 0.95
$contrasts
contrast estimate
                        SE df t.ratio p.value
10 - 20 -2.9111111 0.5934831 24 -4.905 0.0003
10 - 30 -4.3000000 0.5934831 24 -7.245 <.0001
10 - 40 -2.4333333 0.5934831 24 -4.100 0.0022
20 - 30 -1.3888889 0.5934831 24 -2.340 0.1169
30 - 40 1.8666667 0.5934831 24 3.145 0.0213
Results are averaged over the levels of: Variety
```

P value adjustment: tukey method for a family of 4 means

lsmeans(m, pairwise~Variety*Density)

```
$1smeans
Variety Density
                                  SE df lower.CL upper.CL
                    lsmean
                 16.300000 0.7268654 24 14.799824 17.80018
 Α
         10
                  9.200000 0.7268654 24 7.699824 10.70018
         10
                  8.933333 0.7268654 24 7.433157 10.43351
         20
                 18.100000 0.7268654 24 16.599824 19.60018
         20
                 12.433333 0.7268654 24 10.933157 13.93351
                 12.633333 0.7268654 24 11.133157 14.13351
                 19.933333 0.7268654 24 18.433157 21.43351
         30
                 12.900000 0.7268654 24 11.399824 14.40018
         30
         30
                 14.500000 0.7268654 24 12.999824 16.00018
                 18.166667 0.7268654 24 16.666490 19.66684
         40
         40
                 10.800000 0.7268654 24 9.299824 12.30018
 R
         40
                 12.766667 0.7268654 24 11.266490 14.26684
```

Confidence level used: 0.95

\$contrasts

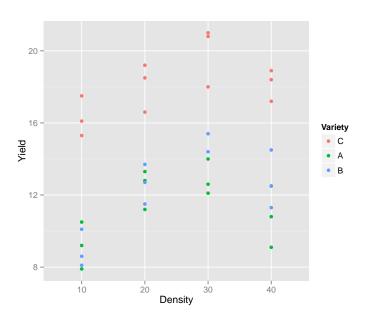
```
contrast
               estimate
                              SE df t.ratio p.value
C,10 - A,10
             7.10000000 1.027943 24
                                     6.907 <.0001
                                     7.166 <.0001
C,10 - B,10 7.36666667 1.027943 24
C,10 - C,20 -1.80000000 1.027943 24
                                     -1.751 0.8276
C,10 - A,20 3.86666667 1.027943 24
                                     3.762 0.0356
C.10 - B.20 3.66666667 1.027943 24
                                     3.567 0.0543
C,10 - C,30 -3.63333333 1.027943 24
                                     -3.535 0.0582
             3.40000000 1.027943 24
                                     3.308
C,10 - A,30
                                            0.0932
C.10 - B.30
             1.80000000 1.027943 24
                                     1.751
                                            0.8276
C,10 - C,40 -1.86666667 1.027943 24
                                     -1.816 0.7947
C,10 - A,40 5.50000000 1.027943 24
                                     5.350
                                            0.0008
C,10 - B,40 3.53333333 1.027943 24
                                     3.437
                                            0.0714
A.10 - B.10 0.26666667 1.027943 24
                                     0.259
                                           1.0000
                                     -8.658
A,10 - C,20 -8.90000000 1.027943 24
                                           < .0001
A.10 - A.20 -3.23333333 1.027943 24
                                     -3.145
                                           0.1284
```

Summary

- Use LSMEANS to answer questions of scientific interest.
- Check model assumptions
- Consider alternative models, e.g. treating density as continuous

Unbalanced design

Suppose for some reason that a variety B, density 30 sample was contaminated. Although you started with a balanced design, the data is now unbalanced. Fortunately, we can still use the tools we have used previously.



Summary statistics

Number of replicates

Mean Yield

```
Variety 10 20 30 40
1 C 16.300000 18.10000 19.93333 18.16667
2 A 9.2000000 12.43333 12.90000 10.80000
3 B 8.933333 12.63333 14.90000 12.76667
```

Standard deviation of yield

```
Variety 10 20 30 40

1 C 1.113553 1.345362 1.6772994 0.8736895

2 A 1.300000 1.096966 0.9848858 1.7000000

3 B 1.040833 1.101514 0.7071068 1.6165808
```

```
DATA tomato;
 INFILE 'Ch13-tomato.csv' DSD FIRSTOBS=2;
 INPUT variety $ density yield;
 i = _n_;
PROC GLM DATA=tomato PLOTS=all:
 WHERE i ~= 19; /* not equal to 19 */
 CLASS variety density;
 MODEL yield = variety|density / SOLUTION;
 LSMEANS variety / cl adjust=tukey;
 LSMEANS density / cl adjust=tukey;
 LSMEANS variety*density / cl adjust=tukey;
 RUN:
```

The GLM Procedure

Dependent Variable: yield

Source Model Error Corrected Total		DF 11 23 34	Sum Squar 423.23885 37.08000 460.31885	res N 571 000	Mean Square 38.4762597 1.6121739	F Value 23.87	Pr > F <.0001
	R-Square	Coef	f Var	Root MSE	E yield M	ean	
	0.919447	9.1	38391	1.269714	13.89	429	
Source		DF	Type I	SS N	Mean Square	F Value	Pr > F
variety		2	329.98787	723 1	164.9939361	102.34	<.0001
density		3	84.44866	808	28.1495536	17.46	<.0001
variety*density		6	8.80232	241	1.4670540	0.91	0.5052
•							
Source		DF	Type III	SS N	Mean Square	F Value	Pr > F
variety		2	320.03746	579 1	160.0187340	99.26	<.0001
density		3	86.06576	313	28.6885871	17.79	<.0001
variety*density		6	8.80232	241	1.4670540	0.91	0.5052

			Standard		
Parameter		Estimate	Error	t Value	Pr > t
Intercept		18.16666667 B	0.73306978	24.78	<.0001
variety	A	-7.36666667 B	1.03671723	-7.11	<.0001
variety	В	-5.40000000 B	1.03671723	-5.21	<.0001
variety	C	0.00000000 B			
density	10	-1.86666667 B	1.03671723	-1.80	0.0849
density	20	-0.06666667 B	1.03671723	-0.06	0.9493
density	30	1.76666667 B	1.03671723	1.70	0.1018
density	40	0.00000000 B	1.000/1/20		
variety*density		0.26666667 B	1.46613956	0.18	0.8573
variety*density		1.70000000 B	1.46613956	1.16	0.2581
, ,					
variety*density		0.33333333 B	1.46613956	0.23	0.8222
variety*density		0.00000000 B	•	•	•
variety*density	B 10	-1.96666667 B	1.46613956	-1.34	0.1929
variety*density	B 20	-0.06666667 B	1.46613956	-0.05	0.9641
variety*density	B 30	0.36666667 B	1.55507584	0.24	0.8157
variety*density	B 40	0.00000000 B			
variety*density	C 10	0.00000000 B			
variety*density	C 20	0.00000000 B			
variety*density		0.00000000 B			
variety*density		0.00000000 B			

```
The GLM Procedure
               Least Squares Means
Adjustment for Multiple Comparisons: Tukey-Kramer
```

Least Squares Means for effect variety Pr > |t| for HO: LSMean(i)=LSMean(j)

Dependent Variable: yield

i/j	1	2	3
1		0.1839	<.0001
2	0.1839		<.0001
3	<.0001	<.0001	
variety	yield LSMEAN	95% Confiden	ce Limits
A	11.333333	10.575098	12.091569
В	12.308333	11.504103	13.112563
C	18.125000	17.366765	18.883235

Least Squares Means for Effect variety

		Difference	Simultane	ous 95%
		Between	Confidence L	imits for
i	j	Means	LSMean(i)-L	SMean(j)
1	2	-0.975000	-2.313097	0.363097
1	3	-6.791667	-8.089811	-5.493522
2	3	-5.816667	-7.154763	-4.478570

The GLM Procedure Least Squares Means

Adjustment for Multiple Comparisons: Tukey-Kramer

Least Squares Means for effect density Pr > |t| for HO: LSMean(i)=LSMean(j)

Dependent Variable: yield

1	2	3	4
	0.0004	<.0001	0.0025
0.0004		0.0967	0.8545
<.0001	0.0967		0.0189
0.0025	0.8545	0.0189	
	<.0001	0.0004 <.0001 0.0967	0.0004 <.0001 0.0004 0.0967 <.0001 0.0967

density	yield LSMEAN	95% Confidence Limits	
10	11.477778	10.602243	12.353312
20	14.388889	13.513354	15.264423
30	15.911111	14.965426	16.856797
40	13.911111	13.035577	14.786646

Least Squares Means for Effect density

		Difference	Simultaneous 95%	
		Between	Confidence L	imits for
i	j	Means	LSMean(i)-L	SMean(j)
1	2	-2.911111	-4.567433	-1.254789
1	3	-4.433333	-6.157288	-2.709379
1	4	-2.433333	-4.089656	-0.777011
2	3	-1.522222	-3.246177	0.201733

The GLM Procedure Least Squares Means

Adjustment for Multiple Comparisons: Tukey-Kramer

			LSMEAN
variety	density	yield LSMEAN	Number
A	10	9.2000000	1
A	20	12.4333333	2
A	30	12.9000000	3
A	40	10.8000000	4
В	10	8.9333333	5
В	20	12.6333333	6
В	30	14.9000000	7
В	40	12.7666667	8
C	10	16.3000000	9
C	20	18.1000000	10
C	30	19.9333333	11
C	40	18.1666667	12

The GLM Procedure
Least Squares Means
Adjustment for Multiple Comparisons: Tukey-Kramer

Least Squares Means for Effect variety*density

		Difference	Simultaneous 95%	
		Between	Confidence	Limits for
i	j	Means	LSMean(i)-	·LSMean(j)
1	11	-10.733333	-14.487164	-6.979502
1	12	-8.966667	-12.720498	-5.212836
2	3	-0.466667	-4.220498	3.287164
2	4	1.633333	-2.120498	5.387164
2	5	3.500000	-0.253831	7.253831
2	6	-0.200000	-3.953831	3.553831
2	7	-2.466667	-6.663577	1.730244
2	8	-0.333333	-4.087164	3.420498
2	9	-3.866667	-7.620498	-0.112836
2	10	-5.666667	-9.420498	-1.912836
2	11	-7.500000	-11.253831	-3.746169
2	12	-5.733333	-9.487164	-1.979502
3	4	2.100000	-1.653831	5.853831
3	5	3.966667	0.212836	7.720498
3	6	0.266667	-3.487164	4.020498
3	7	-2.000000	-6.196911	2.196911
3	8	0.133333	-3.620498	3.887164
3	9	-3.400000	-7.153831	0.353831
3	10	-5.200000	-8.953831	-1.446169
3	11	-7.033333	-10.787164	-3.279502
3	12	-5.266667	-9.020498	-1.512836
4	5	1.866667	-1.887164	5.620498

```
m = lm(Yield~Variety*Density, tomato)
anova(m)
Analysis of Variance Table
Response: Yield
              Df Sum Sq Mean Sq F value Pr(>F)
Variety
              2 327.60 163.799 103.3430 1.608e-12 ***
Density
              3 86.69 28.896 18.2306 2.212e-06 ***
Variety:Density 6 8.03 1.339 0.8445
                                          0.5484
Residuals 24 38.04 1.585
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
lsmeans(m, pairwise~Variety)
$1smeans
 Variety lsmean
                      SE df lower.CL upper.CL
        18.12500 0.3634327 24 17.37491 18.87509
        11.33333 0.3634327 24 10.58325 12.08342
        12.20833 0.3634327 24 11.45825 12.95842
Results are averaged over the levels of: Density
Confidence level used: 0.95
$contrasts
 contrast estimate
                         SE df t.ratio p.value
 C - A 6.791667 0.5139715 24 13.214 <.0001
 C - B 5.916667 0.5139715 24 11.512 <.0001
 A - B -0.875000 0.5139715 24 -1.702 0.2249
Results are averaged over the levels of: Density
P value adjustment: tukey method for a family of 3 means
```

```
lsmeans(m, pairwise Density)
$1smeans
Density lsmean
                     SE df lower.CL upper.CL
       11.47778 0.4196559 24 10.61165 12.34391
10
20
       14.38889 0.4196559 24 13.52276 15.25502
30
       15.77778 0.4196559 24 14.91165 16.64391
40
       13.91111 0.4196559 24 13.04498 14.77724
Results are averaged over the levels of: Variety
Confidence level used: 0.95
$contrasts
contrast estimate
                        SE df t.ratio p.value
10 - 20 -2.9111111 0.5934831 24 -4.905 0.0003
10 - 30 -4.3000000 0.5934831 24 -7.245 <.0001
10 - 40 -2.4333333 0.5934831 24 -4.100 0.0022
20 - 30 -1.3888889 0.5934831 24 -2.340 0.1169
30 - 40 1.8666667 0.5934831 24 3.145 0.0213
Results are averaged over the levels of: Variety
```

P value adjustment: tukey method for a family of 4 means

lsmeans(m, pairwise~Variety*Density)

```
$1smeans
Variety Density
                                  SE df lower.CL upper.CL
                    lsmean
                16.300000 0.7268654 24 14.799824 17.80018
         10
 Α
         10
                  9.200000 0.7268654 24 7.699824 10.70018
         10
                  8.933333 0.7268654 24 7.433157 10.43351
         20
                18.100000 0.7268654 24 16.599824 19.60018
         20
                12.433333 0.7268654 24 10.933157 13.93351
                12.633333 0.7268654 24 11.133157 14.13351
                19.933333 0.7268654 24 18.433157 21.43351
         30
                12.900000 0.7268654 24 11.399824 14.40018
         30
         30
                14.500000 0.7268654 24 12.999824 16.00018
                18.166667 0.7268654 24 16.666490 19.66684
         40
         40
                10.800000 0.7268654 24 9.299824 12.30018
```

12.766667 0.7268654 24 11.266490 14.26684

Confidence level used: 0.95

40

\$contrasts

R

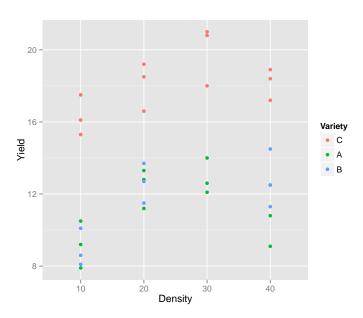
```
contrast
               estimate
                              SE df t.ratio p.value
C,10 - A,10
             7.10000000 1.027943 24
                                     6.907 <.0001
                                     7.166 <.0001
C,10 - B,10 7.36666667 1.027943 24
C,10 - C,20 -1.80000000 1.027943 24
                                    -1.751 0.8276
C,10 - A,20 3.86666667 1.027943 24
                                     3.762 0.0356
C.10 - B.20 3.66666667 1.027943 24
                                     3.567 0.0543
C,10 - C,30 -3.63333333 1.027943 24
                                    -3.535 0.0582
             3.40000000 1.027943 24
                                     3.308
C,10 - A,30
                                           0.0932
C.10 - B.30
             1.80000000 1.027943 24
                                     1.751
                                            0.8276
C,10 - C,40 -1.86666667 1.027943 24
                                    -1.816 0.7947
C,10 - A,40 5.50000000 1.027943 24
                                     5.350
                                            0.0008
C,10 - B,40 3.53333333 1.027943 24
                                     3.437
                                           0.0714
A.10 - B.10 0.26666667 1.027943 24
                                     0.259
                                           1.0000
                                    -8.658
A,10 - C,20 -8.90000000 1.027943 24
                                           < .0001
A.10 - A.20 -3.23333333 1.027943 24
                                    -3.145
                                           0.1284
```

Summary

The analysis can be completed just like the balanced design using LSMEANS to answer scientific questions of interest.

Incomplete design

Suppose none of the samples from Variety B, density 30 were obtained. Now the analysis becomes more complicated.



Summary statistics

Number of replicates

Mean Yield

```
Variety 10 20 30 40
1 C 16.300000 18.10000 19.93333 18.16667
2 A 9.2000000 12.43333 12.90000 10.80000
3 B 8.933333 12.63333 Nan 12.76667
```

Standard deviation of yield

```
Variety 10 20 30 40

1 C 1.113553 1.345362 1.6772994 0.8736895

2 A 1.300000 1.096966 0.984858 1.7000000

3 B 1.040833 1.101514 NA 1.6165808
```

```
INFILE 'Ch13-tomato.csv' DSD FIRSTOBS=2;
INPUT variety $ density yield;

PROC GLM DATA=tomato PLOTS=all;
WHERE ~(variety='B' & density=30);
CLASS variety density;
MODEL yield = variety|density / SOLUTION;
LSMEANS variety / cl adjust=tukey;
LSMEANS density / cl adjust=tukey;
LSMEANS variety*density / cl adjust=tukey;
RUN;
```

DATA tomato:

The GLM Procedure

Dependent	Variable	: yield
-----------	----------	---------

Source Model Error Corrected Total	-	DF 10 22 32	Squa 421.0933 36.5800 457.6733	333 000	Mean Square 42.1093333 1.6627273	F Value 25.33	Pr > F <.0001
	R-Square 0.920074		ff Var 321454	Root MS 1.28946	,		
Source variety density variety*density	,	DF 2 3 5	Type I 347.3819 66.6531 7.0582	444 019	Mean Square 173.6909722 22.2177006 1.4116574	F Value 104.46 13.36 0.85	Pr > F <.0001 <.0001 0.5300
Source variety		DF 2	Type III 321.2233		Mean Square 160.6116898	F Value 96.60	Pr > F <.0001

66.6531019

7.0582870

density

variety*density

22.2177006

1.4116574

13.36

0.85

<.0001

0.5300

Parameter				Standard		
variety A -7.36666667 B 1.05284607 -7.00 <.0001 variety B -5.40000000 B 1.05284607 -5.13 <.0001	Parameter		Estimate	Error	t Value	Pr > t
variety B	Intercept		18.16666667 B	0.74447460	24.40	<.0001
variety C 0.00000000 B 1.05284607 -1.77 0.0901 density 10 -1.86666667 B 1.05284607 -0.06 0.9501 density 20 -0.06666667 B 1.05284607 1.68 0.1075 density 30 1.76666667 B 1.05284607 1.68 0.1075 density 40 0.00000000 B variety*density A 10 0.26666667 B 1.48894919 0.18 0.8595 variety*density A 20 1.70000000 B 1.48894919 1.14 0.2658 variety*density A 30 0.33333333 B 1.48894919 0.22 0.8249 variety*density A 40 0.00000000 B variety*density B 10 -1.96666667 B 1.48894919 -1.32 0.2001 variety*density B 40 0.00000000 B variety*density B 40 0.00000000 B variety*density B 40 0.00000000 B variety*density C 10 0.00000000 B variety*density C 20 0.00000000 B	variety	A	-7.36666667 B	1.05284607	-7.00	<.0001
density 10 -1.86666667 B 1.05284607 -1.77 0.0901 density 20 -0.06666667 B 1.05284607 -0.06 0.9501 density 30 1.76666667 B 1.05284607 -0.06 0.9501 density 40 0.00000000 B	variety	В	-5.40000000 B	1.05284607	-5.13	<.0001
density 20 -0.06666667 B 1.05284607 -0.06 0.9501 density 30 1.766666667 B 1.05284607 1.68 0.1075 density 40 0.0000000 B . variety*density A 10 0.26666667 B 1.48894919 0.18 0.8595 variety*density A 30 0.33333333 B 1.48894919 1.14 0.2658 variety*density A 40 0.0000000 B . variety*density B 10 -1.96666667 B 1.48894919 -1.32 0.201 variety*density B 20 -0.06666667 B 1.48894919 -0.04 0.9647 variety*density B 40 0.0000000 B . variety*density B 40 0.0000000 B . variety*density C 00 0.0000000 B . variety*density C 00 0.0000000 B . variety*density C 00 0.0000000 B .	variety	C	0.00000000 B			
density 30 1.76666667 B 1.05284607 1.68 0.1075 density 40 0.00000000 B	density	10	-1.86666667 B	1.05284607	-1.77	0.0901
density 40 0.0000000 B variety*density A 10 0.26666667 B 1.48894919 0.18 0.8595 variety*density A 20 1.70000000 B 1.48894919 1.14 0.2658 variety*density A 30 0.3333333 B 1.48894919 0.22 0.8249 variety*density B 10 -1.96666667 B 1.48894919 -1.32 0.2001 variety*density B 20 -0.06666667 B 1.48894919 -0.04 0.9647 variety*density B 40 0.00000000 B variety*density B 40 0.00000000 B variety*density C 10 0.00000000 B variety*density C 20 0.00000000 B	density	20	-0.06666667 B	1.05284607	-0.06	0.9501
<pre>variety*density A 10</pre>	density	30	1.76666667 B	1.05284607	1.68	0.1075
<pre>variety*density A 20</pre>	density	40	0.00000000 B			
variety*density A 30	variety*density	A 10	0.26666667 B	1.48894919	0.18	0.8595
<pre>variety*density A 40</pre>	variety*density	A 20	1.70000000 B	1.48894919	1.14	0.2658
<pre>variety*density B 10</pre>	variety*density	A 30	0.33333333 B	1.48894919	0.22	0.8249
<pre>variety*density B 20</pre>	variety*density	A 40	0.00000000 B			
<pre>variety*density B 40</pre>	variety*density	B 10	-1.96666667 B	1.48894919	-1.32	0.2001
<pre>variety*density C 10</pre>	variety*density	B 20	-0.06666667 B	1.48894919	-0.04	0.9647
variety*density C 20 0.00000000 B	variety*density	B 40	0.00000000 B			
	variety*density	C 10	0.00000000 B			
variety*density C 30	variety*density	C 20	0.00000000 B			
railoup admitty of the control of th	variety*density	C 30	0.00000000 B			
variety*density C 40 0.00000000 B	variety*density	C 40	0.00000000 B	•		

Notice the missing variety*density B 30 line.

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

		LSMEAN
variety	yield LSMEAN	Number
A	11.3333333	1
В	Non-est	2
C	18.1250000	3

Least Squares Means for effect variety
Pr > |t| for HO: LSMean(i)=LSMean(j)

Least Squares Means for Effect variety

		Difference	Simultane	ous 95%
		Between	Confidence L	imits for
i	j	Means	LSMean(i)-L	SMean(j)
1	2			
1	3	-6.791667	-7.883358	-5.69997
2	3			

5

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

		LSMEAN
density	yield LSMEAN	Number
10	11.4777778	1
20	14.3888889	2
30	Non-est	3
40	13.9111111	4

density	yield LSMEAN	95% Confiden	ce Limits
10	11.477778	10.586380	12.369175
20	14.388889	13.497491	15.280286
30			
40	13.911111	13.019714	14.802509

Least Squares Means for Effect density

		Difference	Simultane	ous 95%
		Between	Confidence L	imits for
i	j	Means	LSMean(i)-L	SMean(j)
1	2	-2.911111	-4.438096	-1.384126
1	3			
1	4	-2.433333	-3.960319	-0.906348
2	3			
2	4	0.477778	-1.049207	2.004763
3	4			

The GLM Procedure

Least Squares Means
Adjustment for Multiple Comparisons: Tukey

variety	density	yield LSMEAN	LSMEAN Number
A	10	9.2000000	1
A	20	12.4333333	2
A	30	12.9000000	3
A	40	10.8000000	4
В	10	8.9333333	5
В	20	12.6333333	6
В	40	12.7666667	7
C	10	16.3000000	8
C	20	18.1000000	9
C	30	19.9333333	10

		Difference	Simultan	eous 95%
		Between	Confidence	Limits for
i	j	Means	LSMean(i)-	LSMean(j)
1	2	-3.233333	-6.997053	0.530387
1	3	-3.700000	-7.463720	0.063720
1	4	-1.600000	-5.363720	2.163720
1	5	0.266667	-3.497053	4.030387
1	6	-3.433333	-7.197053	0.330387
1	7	-3.566667	-7.330387	0.197053
1	8	-7.100000	-10.863720	-3.336280
1	9	-8.900000	-12.663720	-5.136280
1	10	-10.733333	-14.497053	-6.969613
1	11	-8.966667	-12.730387	-5.202947
2	3	-0.466667	-4.230387	3.297053
2	4	1.633333	-2.130387	5.397053
2	5	3.500000	-0.263720	7.263720
2	6	-0.200000	-3.963720	3.563720
2	7	-0.333333	-4.097053	3.430387
2	8	-3.866667	-7.630387	-0.102947
2	9	-5.666667	-9.430387	-1.902947
2	10	-7.500000	-11.263720	-3.736280
2	11	-5.733333	-9.497053	-1.969613
3	4	2.100000	-1.663720	5.863720
3	5	3.966667	0.202947	7.730387
3	6	0.266667	-3.497053	4.030387
3	7	0.133333	-3.630387	3.897053
3	8	-3.400000	-7.163720	0.363720
3	9	-5.200000	-8.963720	-1.436280
3	10	-7.033333	-10.797053	-3.269613
3	11	-5.266667	-9.030387	-1.502947
4	5	1.866667	-1.897053	5.630387

Treat as a One-way ANOVA

When the data are incomplete, use a one-way ANOVA combined with contrasts to answer questions of interest. For example, to compare the average difference between B and C, we want to only compare at densities 10, 20, and 40.

	10	20	30	40
Α	μ_{11}	μ_{12}	μ_{13}	μ_{14}
В	μ_{21}	μ_{22}	μ_{23}	μ_{24}
С	μ_{31}	μ_{32}	μ_{33}	μ 34

Thus, the contrast is

$$\gamma = \frac{1}{3}(\mu_{31} + \mu_{32} + \mu_{34}) - \frac{1}{3}(\mu_{21} + \mu_{22} + \mu_{24})
= \frac{1}{3}(\mu_{31} + \mu_{32} + \mu_{34} - \mu_{21} - \mu_{22} - \mu_{24})$$

```
DATA tomato:
 INFILE 'Ch13-tomato.csv' DSD FIRSTOBS=2:
 INPUT variety $ density yield;
PROC GLM DATA=tomato PLOTS=all:
 WHERE ~(variety='B' & density=30);
 CLASS variety density;
 MODEL vield = variety*density / SOLUTION CLPARM;
 LSMEANS variety*density / cl adjust=tukey;
 /*
                               A10 A20 A30 A40 B10 B20 B40 C10 C20 C30 C40 */
 ESTIMATE 'C-A' variety*density -1 -1 -1 -1 0 0 0 1 1 1 1/DIVISOR-4; ESTIMATE 'B-A' variety*density -1 -1 0 -1 1 1 1 0 0 0 0 0 JUVISOR-3;
 /* we could do the densities similarly */
 RUN:
```

The GLM Procedure

Dependent Variable: yi	ета
------------------------	-----

		S	um of			
Source	DF	Sq	uares	Mean Square	F Value	Pr > F
Model	10	421.09	33333	42.1093333	25.33	<.0001
Error	22	36.58	00000	1.6627273		
Corrected Total	32	457.67	33333			
R-Square	Coe	ff Var	Root	MSE yield M	lean	
0.920074	9.	321454	1.28	9468 13.83	3333	
Source	DF	Туре	I SS	Mean Square	F Value	Pr > F
variety*density	10	421.09	33333	42.1093333	25.33	<.0001
Source	DF	Type I	II SS	Mean Square	F Value	Pr > F
variety*density	10	421.09	33333	42.1093333	25.33	<.0001

		Standard				
Parameter	Estimate	Error	t Value	Pr > t	95% Confide	nce Limits
Intercept	18.16666667 B	0.74447460	24.40	<.0001	16.62272085	19.71061248
variety*density A 10	-8.96666667 B	1.05284607	-8.52	<.0001	-11.15013578	-6.78319756
variety*density A 20	-5.73333333 B	1.05284607	-5.45	<.0001	-7.91680244	-3.54986422
variety*density A 30	-5.26666667 B	1.05284607	-5.00	<.0001	-7.45013578	-3.08319756
variety*density A 40	-7.36666667 B	1.05284607	-7.00	<.0001	-9.55013578	-5.18319756
variety*density B 10	-9.23333333 B	1.05284607	-8.77	<.0001	-11.41680244	-7.04986422
variety*density B 20	-5.53333333 B	1.05284607	-5.26	<.0001	-7.71680244	-3.34986422
variety*density B 40	-5.40000000 B	1.05284607	-5.13	<.0001	-7.58346911	-3.21653089
variety*density C 10	-1.86666667 B	1.05284607	-1.77	0.0901	-4.05013578	0.31680244
variety*density C 20	-0.06666667 B	1.05284607	-0.06	0.9501	-2.25013578	2.11680244
variety*density C 30	1.76666667 B	1.05284607	1.68	0.1075	-0.41680244	3.95013578
varietv*densitv C 40	0.00000000 B					

The Regression model

The regression model here considers variety-density combination as a single explanatory variable with 11 levels: A10, A20, A30, A40, B10, B20, B40, C10, C20, C30, and C40. By default, SAS chose C40 as our reference level. For observation i, let

- Y_i be the yield
- \bullet V_i be the variety
- D_i be the density

The model is then $Y_i \stackrel{ind}{\sim} N(\mu_i, \sigma^2)$ and

The GLM Procedure

Dependent Variable: yield

Standard

Parameter	Estimate	Error	t Value	Pr > t	95% Confider	nce Limits
C-B	6.07777778	0.60786096	10.00	<.0001	4.81715130	7.33840426
C-A	6.79166667	0.52642304	12.90	<.0001	5.69993211	7.88340122
B-A	0.63333333	0.60786096	1.04	0.3088	-0.62729315	1.89395981

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey

			LSMEAN
variety	density	yield LSMEAN	Number
A	10	9.2000000	1
A	20	12.4333333	2
A	30	12.9000000	3
A	40	10.8000000	4
В	10	8.9333333	5
В	20	12.6333333	6
В	40	12.7666667	7
C	10	16.3000000	8
C	20	18.1000000	9
C	30	19.9333333	10

		Difference	Simultane	ous 95%
		Between	Confidence L	imits for
i	j	Means	LSMean(i)-L	SMean(j)
1	2	-3.233333	-6.997053	0.530387
1	3	-3.700000	-7.463720	0.063720
1	4	-1.600000	-5.363720	2.163720
1	5	0.266667	-3.497053	4.030387
1	6	-3.433333	-7.197053	0.330387
1	7	-3.566667	-7.330387	0.197053
1	8	-7.100000	-10.863720	-3.336280
1	9	-8.900000	-12.663720	-5.136280
1	10	-10.733333	-14.497053	-6.969613
1	11	-8.966667	-12.730387	-5.202947
2	3	-0.466667	-4.230387	3.297053
2	4	1.633333	-2.130387	5.397053
2	5	3.500000	-0.263720	7.263720
2	6	-0.200000	-3.963720	3.563720
2	7	-0.333333	-4.097053	3.430387
2	8	-3.866667	-7.630387	-0.102947
2	9	-5.666667	-9.430387	-1.902947
2	10	-7.500000	-11.263720	-3.736280
2	11	-5.733333	-9.497053	-1.969613
3	4	2.100000	-1.663720	5.863720
3	5	3.966667	0.202947	7.730387
3	6	0.266667	-3.497053	4.030387
3	7	0.133333	-3.630387	3.897053
3	8	-3.400000	-7.163720	0.363720
3	9	-5.200000	-8.963720	-1.436280
3	10	-7.033333	-10.797053	-3.269613
3	11	-5.266667	-9.030387	-1.502947
4	5	1.866667	-1.897053	5.630387

Analysis in R

```
m = lm(Yield~Variety:Density, tomato, subset=!(Variety=='B' & Density==30))
anova(m)
Analysis of Variance Table
Response: Yield
               Df Sum Sq Mean Sq F value Pr(>F)
Variety:Density 10 421.09 42.109 25.326 8.563e-10 ***
Residuals
           22 36 58 1 663
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
tomato$VarietyDensity = factor(paste(tomato$Variety, tomato$Density, sep=""))
# Note the -1 in order to construct the contrast
m = lm(Yield~VarietyDensity-1, tomato, subset=!(Variety=='B' & Density==30))
                   A10 A20 A30 A40 B10 B20 B40 C10 C20 C30 C40
K = rbind('C-B' = c(0, 0, 0, 0, -1, -1, -1, 1, 1, 0, 1)/3,
          ^{\prime}C-A^{\prime} = c(-1, -1, -1, -1, 0, 0, 0, 1, 1, 1, 1)/4,
          'B-A' = c(-1, -1, 0, -1, 1, 1, 1, 0, 0, 0, 0)/3)
library(multcomp)
t = glht(m, linfct=K)
#summary(t)
confint(t, calpha=univariate calpha())
 Simultaneous Confidence Intervals
Fit: lm(formula = Yield ~ VarietyDensity - 1, data = tomato, subset = !(Variety ==
    "B" & Density == 30))
Quantile = 2.0739
95% confidence level
Linear Hypotheses:
         Estimate lwr upr
C-B == 0 6.0778 4.8172 7.3384
C-A == 0 6.7917 5.6999 7.8834
B-A == 0 \quad 0.6333 \quad -0.6273 \quad 1.8940
```

```
m = lm(Yield~Variety:Density, tomato, subset=!(Variety=='B' & Density==30))
lsmeans(m, pairwise~Variety:Density)
```

\$1smeans

```
Variety Density
                   lsmean
                                 SE df lower.CL upper.CL
                16.300000 0.7444746 22 14.756054 17.84395
        10
                 9.200000 0.7444746 22 7.656054 10.74395
        10
       10
                 8.933333 0.7444746 22 7.389388 10.47728
C.
        20
                18.100000 0.7444746 22 16.556054 19.64395
                12.433333 0.7444746 22 10.889388 13.97728
        20
                12.633333 0.7444746 22 11.089388 14.17728
        20
                19.933333 0.7444746 22 18.389388 21.47728
        30
        30
                12.900000 0.7444746 22 11.356054 14.44395
                                 NA NA
        30
                       NA
                                              NA
                                                        NA
        40
                18.166667 0.7444746 22 16.622721 19.71061
        40
                10.800000 0.7444746 22 9.256054 12.34395
R
        40
                12.766667 0.7444746 22 11.222721 14.31061
```

Confidence level used: 0.95

\$contrasts

```
contrast
               estimate
                              SE df t.ratio p.value
C.10 - A.10
             7.10000000 1.052846 22
                                   6.744 <.0001
C,10 - B,10 7.36666667 1.052846 22
                                     6.997 < .0001
C,10 - C,20 -1.80000000 1.052846 22
                                    -1.710 0.8458
C.10 - A.20 3.86666667 1.052846 22
                                     3.673 0.0465
C.10 - B.20 3.66666667 1.052846 22
                                     3.483 0.0688
C,10 - C,30 -3.63333333 1.052846 22
                                    -3.451 0.0734
C.10 - A.30
             3.40000000 1.052846 22
                                     3.229
                                            0.1136
C.10 - B.30
                     NA
                              NA NA
                                        NA
                                                NA
C,10 - C,40 -1.86666667 1.052846 22
                                    -1.773 0.8156
C,10 - A,40 5.50000000 1.052846 22
                                     5.224 0.0014
C.10 - B.40 3.53333333 1.052846 22
                                     3.356
                                           0.0887
             0.26666667 1.052846 22
                                     0.253
A,10 - B,10
                                           1.0000
A,10 - C,20 -8.90000000 1.052846 22
                                    -8.453
                                           <.0001
```

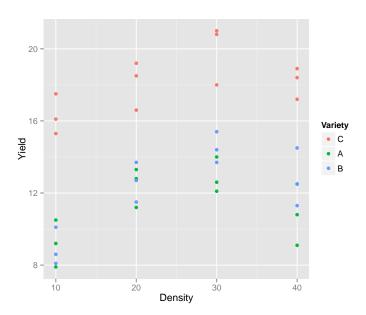
Summary

When dealing with an incomplete design, it is often easier to treat the analysis as a one-way ANOVA and use contrasts to answer scientific questions of interest.

Optimal yield

Now suppose you have the same data set, but your scientific question is different. Specifically, you are interested in choosing a variety and density that provide the optimal yield.

You can use the ANOVA analysis to choose from amongst the 3 varieties and one of the 4 densities, but there is no reason to believe that the optimal density will be one of those 4.



Modeling

Considering a single variety, if we assume a linear relationship between Yield (Y_i) and Density (D_i) then the maximum Yield will occur at either $-\infty$ or $+\infty$ which is unreasonable. The easiest way to have a maximum (or minimum) is to assume a quadratic relationship, e.g.

$$E[Y_i] = \mu_i = \beta_0 + \beta_1 D_i + \beta_2 D_i^2$$

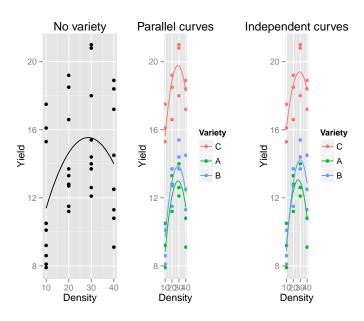
Now we can incorporate Variety (V_i) in many ways. Two options are parallel curves or completely independent curves.

Parallel curves:

$$\mu_i = \beta_0 + \beta_1 D_i + \beta_2 D_i^2 + \beta_3 I(V_i = A) + \beta_4 I(V_i = B)$$

Independent lines:

$$\mu_i = \beta_0 + \beta_1 D_i + \beta_2 D_i^2$$
Two-way ANOVA



Finding the maximum

For a particular variety, there will be an equation like

$$E[Y_i] = \mu_i = \beta_0 + \beta_1 D_i + \beta_2 D_i^2$$

where these β_1 and β_2 need not correspond to any particular β_1 and β_2 we have discussed thus far.

If $\beta_2 < 0$, then the quadratic curve has a maximum and it occurs at $-\beta_1/2\beta_2$.

Analysis in SAS

DATA tomato;

```
INFILE 'Ch13-tomato.csv' DSD FIRSTOBS=2:
 INPUT variety $ density yield;
/* No variety */
PROC GLM DATA=tomato PLOTS=all:
 CLASS variety; /* density is no longer here */
 MODEL yield = density | density / SOLUTION;
 RUN:
/* Parallel curves */
PROC GLM DATA=tomato PLOTS=all:
 CLASS variety; /* density is no longer here */
 MODEL yield = density|density variety/ SOLUTION;
 RUN:
/* Independent curves */
PROC GLM DATA=tomato PLOTS=all;
 CLASS variety: /* density is no longer here */
 MODEL yield = density|density|variety/ SOLUTION;
 RUN;
```

No variety

The GLM Procedure

Dependent Variable: yield

		Sum of			
Source	DF	Squares	Mean Square	F Value	Pr > F
Model	2	85.3346667	42.6673333	3.75	0.0340
Error	33	375.0208889	11.3642694		
Corrected Total	35	460.355556			

Source	DF	Type III SS	Mean Square	F Value	Pr > F
density	1	65.30344358	65.30344358	5.75	0.0223
density*density	1	51.36111111	51.36111111	4.52	0.0411

Standard

Parameter	Estimate	Error	t Value	Pr > t
Intercept	5.74444444	3.12824210	1.84	0.0753
density	0.684111111	0.28538383	2.40	0.0223
density*density	-0.011944444	0.00561849	-2.13	0.0411

Parallel curves

The GLM Procedure

Dependent Variable: yield

		Sum of			
Source	DF	Squares	Mean Square	F Value	Pr > F
Model	4	412.9318889	103.2329722	67.48	<.0001
Error	31	47.4236667	1.5297957		
Corrected Total	35	460.3555556			

Source	DF	Type III SS	Mean Square	F Value	Pr > F
density	1	65.3034436	65.3034436	42.69	<.0001
density*density	1	51.3611111	51.3611111	33.57	<.0001
variety	2	327.5972222	163.7986111	107.07	<.0001

			Standard		
Parameter		Estimate	Error	t Value	Pr > t
Intercept		9.980555556 B	1.18419286	8.43	<.0001
density		0.684111111	0.10470690	6.53	<.0001
density*dens	sity	-0.011944444	0.00206142	-5.79	<.0001
variety	A	-6.791666667 B	0.50494153	-13.45	<.0001
variety	В	-5.916666667 B	0.50494153	-11.72	<.0001
variety	C	0.000000000 B			

Independent curves

			Sum of			
Source		DF	Squares	Mean Squa	re F Val	ue Pr > F
Model		8	419.8612222	52.48265	28 34.	99 <.0001
Error		27	40.4943333	1.49979	01	
Corrected Total		35	460.355556			
Source		DF	Type III SS	Mean Squa	re F Val	ue Pr > F
density		1	65.30344358	65.303443	58 43.	54 <.0001
density*density		1	51.36111111	51.361111	11 34.	25 <.0001
variety		2	21.66539427	10.832697	13 7.	22 0.0031
density*variety		2	2.07850215	1.039251	08 0.	69 0.5088
densit*densit*variet		2	1.65388889	0.826944	44 0.	55 0.5825
				Standard		
Parameter		E	Estimate	Error	t Value	Pr > t
Intercept		11.8	30833333 B	1.96836425	6.00	<.0001
density		0.5	52016667 B	0.17957029	2.90	0.0074
density*density		-0.0	00891667 B	0.00353529	-2.52	0.0179
variety	Α	-8.4	15833333 B	2.78368742	-3.04	0.0052
variety	В	-9.7	73333333 B	2.78368742	-3.50	0.0016
variety	C	0.0	00000000 B			
density*variety	A	0.1	19916667 B	0.25395073	0.78	0.4397
density*variety	В		29266667 B	0.25395073	1.15	0.2592
density*variety	C		00000000 B	•	•	
densit*densit*variet			00441667 B	0.00499965	-0.88	0.3848
densit*densit*variet	-		00466667 B	0.00499965	-0.93	0.3589
densit*densit*variet	C	0.0	00000000 B	•	•	•

No variety

```
Call:
lm(formula = Yield ~ Density + I(Density^2), data = tomato)
Residuals:
  Min 1Q Median 3Q Max
-4.898 -2.721 -1.320 3.364 6.109
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 5.744444 3.128242 1.836 0.0753 .
Density 0.684111 0.285384 2.397 0.0223 *
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
Residual standard error: 3.371 on 33 degrees of freedom
Multiple R-squared: 0.1854, Adjusted R-squared: 0.136
F-statistic: 3.755 on 2 and 33 DF, p-value: 0.03395
```

Parallel curves

```
Call:
lm(formula = Yield ~ Densitv + I(Densitv^2) + Varietv, data = tomato)
Residuals:
   Min
          10 Median 30
                               Max
-2.3422 -0.9039 0.1744 0.8082 2.1828
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 9.980556 1.184193 8.428 1.61e-09 ***
Density
         0.684111 0.104707 6.534 2.71e-07 ***
VarietyA -6.791667 0.504942 -13.450 1.76e-14 ***
VarietyB -5.916667 0.504942 -11.718 6.39e-13 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.237 on 31 degrees of freedom
Multiple R-squared: 0.897, Adjusted R-squared: 0.8837
F-statistic: 67.48 on 4 and 31 DF, p-value: 7.469e-15
```

Independent curves

```
Call:
lm(formula = Yield ~ Density * Variety + I(Density^2) * Variety,
   data = tomato)
Residuals:
    Min
             1Q Median
                              3Q
                                     Max
-2.04500 -0.82125 -0.01417 0.94000 1.71000
Coefficients:
                     Estimate Std. Error t value Pr(>|t|)
(Intercept)
                   11.808333 1.968364 5.999 2.12e-06 ***
                   0.520167 0.179570 2.897 0.00739 **
Density
                   -8.458333 2.783687 -3.039 0.00523 **
VarietvA
VarietyB
                  -9.733333 2.783687 -3.497 0.00165 **
I(Density^2)
                -0.008917
                              0.003535 -2.522 0.01787 *
Density: VarietyA 0.199167
                              0.253951 0.784 0.43971
Density: VarietyB 0.292667
                              0.253951 1.152 0.25924
VarietyA:I(Density^2) -0.004417
                              0.005000 -0.883 0.38482
VarietyB:I(Density^2) -0.004667
                               0.005000 -0.933 0.35889
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.225 on 27 degrees of freedom
Multiple R-squared: 0.912, Adjusted R-squared: 0.886
F-statistic: 34.99 on 8 and 27 DF, p-value: 2.678e-12
```

Completely randomized design (CRD)

This semester, we have assumed a completely randomized design. As an example, consider 36 plots and we are randomly assigning our variety-density combinations to the plots such that we have 3 reps of each combination. The result may look something like this

A20	A30	A40	C20	A40	B40
C20	C40	C40	B30	A10	A40
B40	C30	B40	C10	A20	C10
C10	B20	B20	A30	A30 B10	
A10	C40	A10	B10	A30	B10
C20	B30	B20	B30	C30	C30

Complete randomized block design (RBD)

A randomized block design is appropriate when there is a nuisance factor that you want to control for. In our example, imagine you had 12 plots at 3 different locations and you expect these locations would have impact on yield. A randomized block design might look like this.

A30	B40
C10	B10
C30	C20
B30	B20
A10	A20
C40	A40

A20	B40	
C10	B20	
C30	C40	
A10	A30	
B30	A40	
C20	B10	

A10	B40
C20	B30
C10	A40
A20	C40
A30	B10
B20	C30

Block 1

Block 2

Block 3

RBD Analysis

Generally, you will want to model a randomized block design using an additive model for the treatment and blocking factor. If you have the replication, you should test for an interaction. Let's compute the degrees of freedom for the ANOVA tables for this current design considering the variety-density combination as the treatment.

V+D+B		T+B		Cell-means	
Factor	df	Factor	df	Factor	df
Variety	2				
Density	3	Treatment	11	Treatment	11
Block	2	Block	2	Block	2
				Treatment x Block	22
Error	28	Error	22	Error	0
Total	35	Total	35	Total	35

The cell-means model does not have enough degrees of freedom to estimate the interacion because there is no replication of the treatment within a block.

Why blocking helps

Estimation of treatment effects is performed within block because all treatments are available within a block.

Block what you can; randomize what you cannot.