

Output from bleach.R

```
> # bleach.R
>
> library(lsmmeans)
> library(car)
> library(phia)
> library(doBy)
>
> # get the data
> bleach<-read.table(file="bleach.dat",header=T)
> head(bleach)
  amtblch staintyp time
1      3      ink 3600
2      3      ink 3920
3      3      ink 3340
4      3      ink 3173
5      3      jam  495
6      3      jam  236
> bleach$amtfac <- factor(bleach$amtblch)
> is.ordered(bleach$amtfac)
[1] FALSE
> is.ordered(bleach$staintyp)
[1] FALSE
>
> #####PART 1
>
> # fit the two-way anova model
> m1 <- aov(time~amtfac*staintyp,data=bleach)
> summary(m1)
              Df    Sum Sq Mean Sq F value    Pr(>F)
amtfac          2  1229568   614784    7.662 0.00191 **
staintyp        2 64714567 32357284 403.245 < 2e-16 ***
amtfac:staintyp  4  5411147  1352787  16.859 1.57e-07 ***
Residuals      32  2567752    80242
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> # note that the summary function gives F tests based on type I
> # or sequential sums of squares. To get F tests based on the type II and III
> # SS, we need the Anova() function from the car package.
>
> Anova(m1,type=2)
Anova Table (Type II tests)

Response: time
              Sum Sq Df  F value    Pr(>F)
amtfac          1184748  2    7.3823 0.002311 **
staintyp        64714567  2 403.2449 < 2.2e-16 ***
amtfac:staintyp  5411147  4  16.8588 1.569e-07 ***
Residuals      2567752 32
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> # the Anova() function from the car package claims to produce Type III
> # SSs and F tests, but if you run the line below you'll see that you get
> # results quite different from those provided by SAS.
> Anova(m1,type=3)
```

Anova Table (Type III tests)

Response: time

	Sum Sq	Df	F value	Pr(>F)
(Intercept)	85505806	1	1065.5957	< 2.2e-16 ***
amtfac	917710	2	5.7184	0.007528 **
staintyp	65287167	2	406.8129	< 2.2e-16 ***
amtfac:staintyp	5411147	4	16.8588	1.569e-07 ***
Residuals	2567752	32		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
>
> # The reason for this discrepancy is related to the method of computing
> # type III SSs in the Anova() function. That method depends upon the
> # parameterization of the model. R and SAS parameterize factors differently.
> # SAS uses a "non-full rank" parameterization (aka an overparameterization),
> # whereas R removes the overparameterization. The user has some control over
> # how that overparameterization is removed. That is, the user has some control
> # over the parameterization that R uses. For the Anova(,type=3) function to
> # produce the "right" type III SSs (the ones that agree with SAS and test
> # hypotheses on marginal means as described in our lecture notes), one
> # has to choose a sum-to-zero parameterization of factor effects. This is
> # not the default in R. Therefore it has to be changed and the model refit.
```

```
> # Change the handling of unordered factors (like amtfac and staintyp) to
> # use the sum-to-zero constraints:
```

```
> op <- options(contrasts=c("contr.sum", "contr.poly"))
> options()$contrasts # now you can see that they've been changed
```

```
[1] "contr.sum" "contr.poly"
```

```
>
> # Now refit the model with the new parameterization:
```

```
> mla <- aov(time~amtfac*staintyp,data=bleach)
```

```
> # now check the type 1 SSs (they haven't changed):
```

```
> summary(mla)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
amtfac	2	1229568	614784	7.662	0.00191 **
staintyp	2	64714567	32357284	403.245	< 2e-16 ***
amtfac:staintyp	4	5411147	1352787	16.859	1.57e-07 ***
Residuals	32	2567752	80242		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
>
> # and check the type II SSs (they haven't changed):
```

```
> Anova(mla,type=2)
```

Anova Table (Type II tests)

Response: time

	Sum Sq	Df	F value	Pr(>F)
amtfac	1184748	2	7.3823	0.002311 **
staintyp	64714567	2	403.2449	< 2.2e-16 ***
amtfac:staintyp	5411147	4	16.8588	1.569e-07 ***
Residuals	2567752	32		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
>
> # and check the type III SSs (they have changed and now agree with SAS):
```

```
> Anova(mla,type=3)
```

Anova Table (Type III tests)

Response: time

	Sum Sq	Df	F value	Pr(>F)
(Intercept)	85505806	1	1065.5957	< 2.2e-16 ***
amtfac	917710	2	5.7184	0.007528 **
staintyp	65287167	2	406.8129	< 2.2e-16 ***
amtfac:staintyp	5411147	4	16.8588	1.569e-07 ***
Residuals	2567752	32		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
>
> #####PART 2
>
> # Now get the marginal and joint lsmeans:
> lsmeans(m1, specs = list(~staintyp, ~amtfac, ~amtfac:staintyp))
```

```
$`staintyp` lsmeans`
  staintyp    lsmean      SE df lower.CL upper.CL
    ink 3234.4167 76.12670 32 3079.3517 3389.4817
    jam  385.9833 76.12670 32  230.9183  541.0483
    tomato 774.8222 80.85949 32  610.1168  939.5276
```

```
$`amtfac` lsmeans`
  amtfac    lsmean      SE df lower.CL upper.CL
    3 1566.817 76.12670 32 1411.752 1721.882
    5 1252.650 76.12670 32 1097.585 1407.715
    7 1575.756 80.85949 32 1411.050 1740.461
```

```
$`amtfac:staintyp` lsmeans`
  amtfac staintyp    lsmean      SE df lower.CL upper.CL
    3      ink 3508.2500 141.6353 32 3219.74829 3796.7517
    5      ink 2350.8000 126.6825 32 2092.75623 2608.8438
    7      ink 3844.2000 126.6825 32 3586.15623 4102.2438
    3      jam  474.8000 126.6825 32  216.75623  732.8438
    5      jam  370.7500 141.6353 32   82.24829  659.2517
    7      jam  312.4000 126.6825 32   54.35623  570.4438
    3     tomato 717.4000 126.6825 32  459.35623  975.4438
    5     tomato 1036.4000 126.6825 32  778.35623 1294.4438
    7     tomato  570.6667 163.5464 32  237.53359  903.7997
```

Warning messages:

```
1: In lsmeans(m1, specs = list(~staintyp, ~amtfac, ~amtfac:staintyp)) :
  lsmeans of staintyp may be misleading due to interaction with other
  predictor(s)
2: In lsmeans(m1, specs = list(~staintyp, ~amtfac, ~amtfac:staintyp)) :
  lsmeans of amtfac may be misleading due to interaction with other predictor(s)
>
```

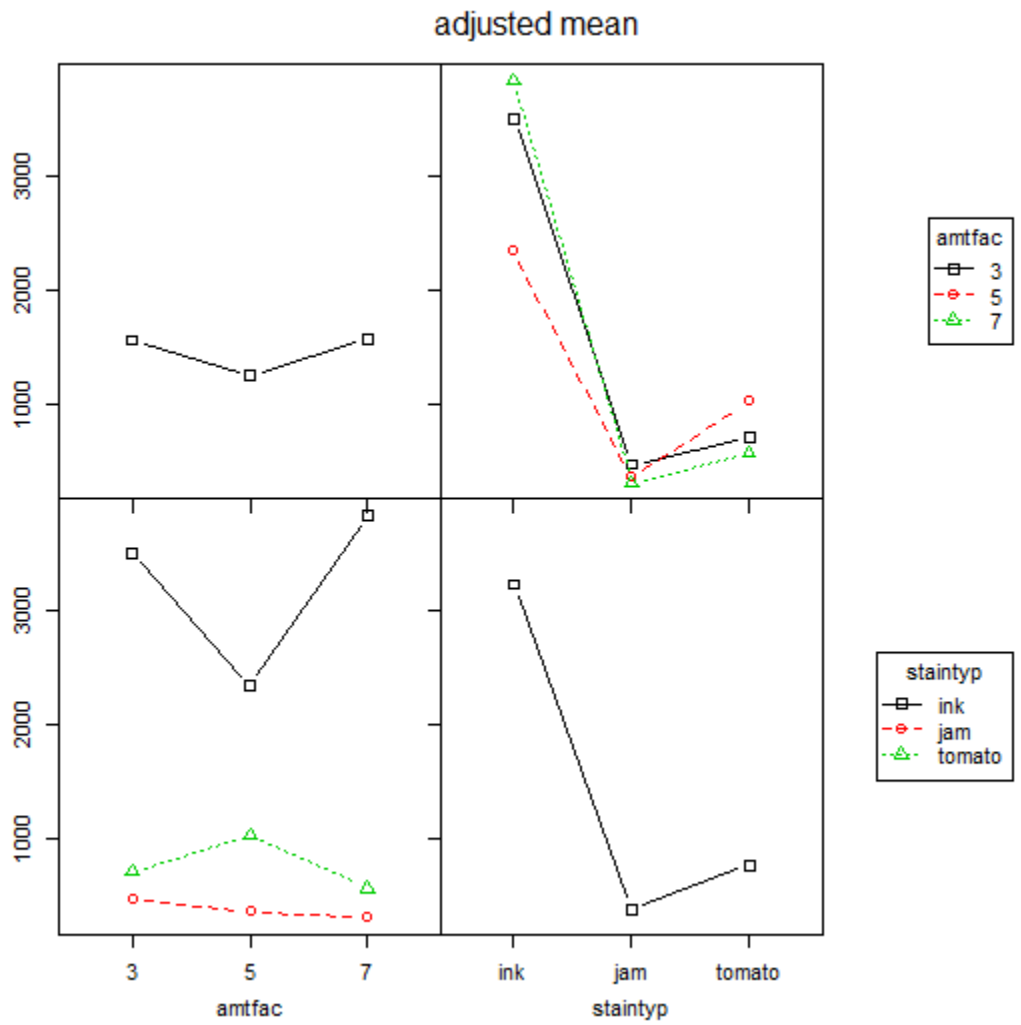
```
> # get profile plots (uses package phia)
> ( trtmns <- interactionMeans(m1) )
```

	amtfac	staintyp	adjusted mean
1	3	ink	3508.2500
2	5	ink	2350.8000
3	7	ink	3844.2000
4	3	jam	474.8000
5	5	jam	370.7500
6	7	jam	312.4000
7	3	tomato	717.4000

```

8      5  tomato  1036.4000
9      7  tomato  570.6667
> plot(trtmns)

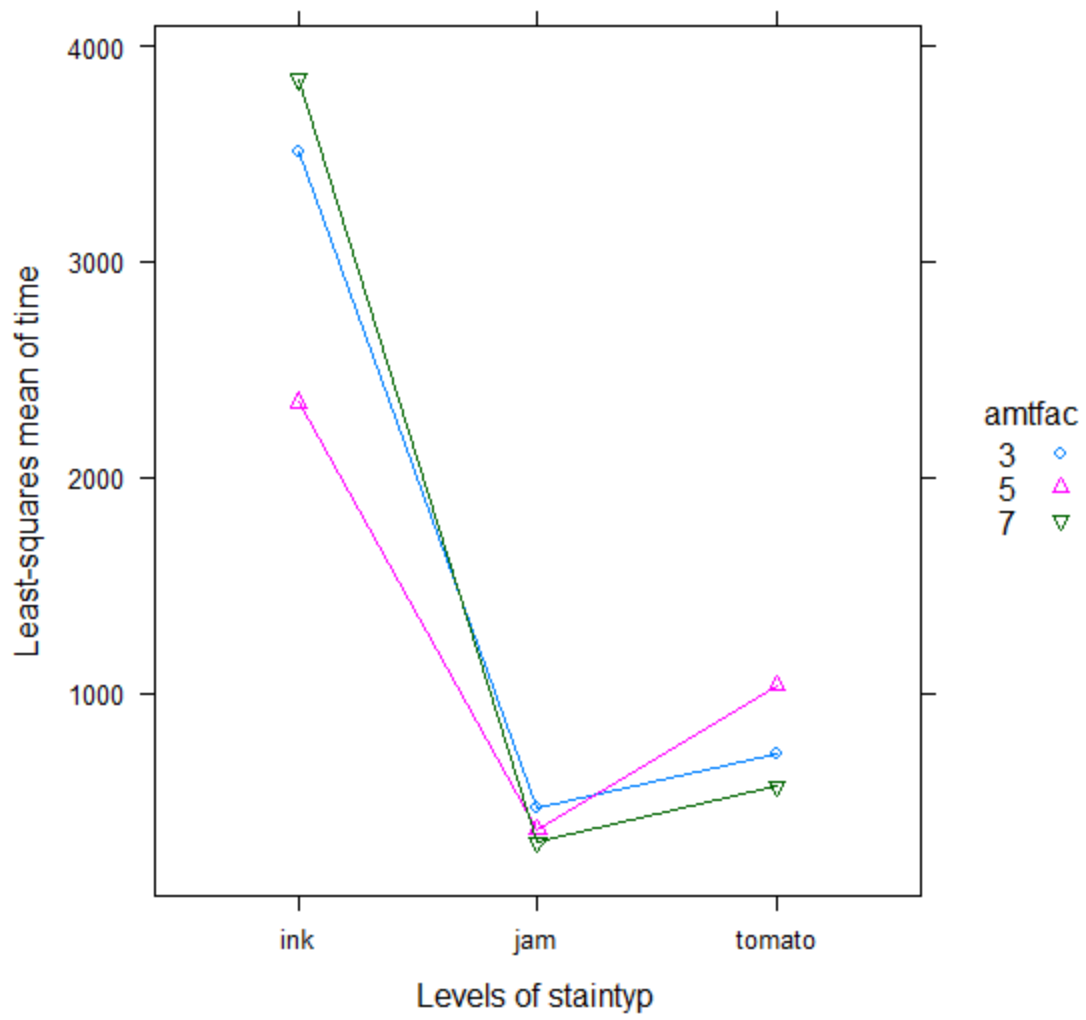
```



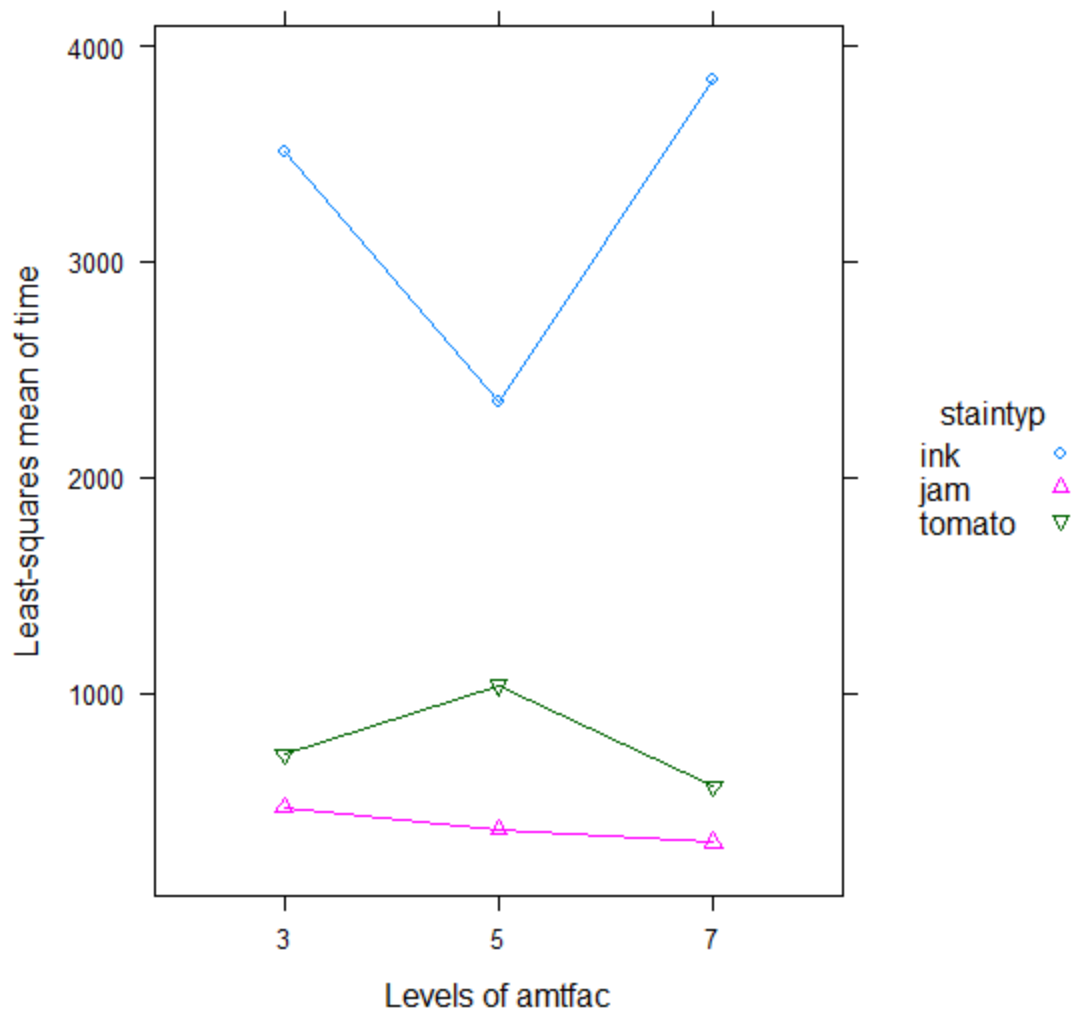
```

>
> # alternatively can use function lsmip from lsmeans package
>
> lsmip(m1, amtfac~staintyp)

```



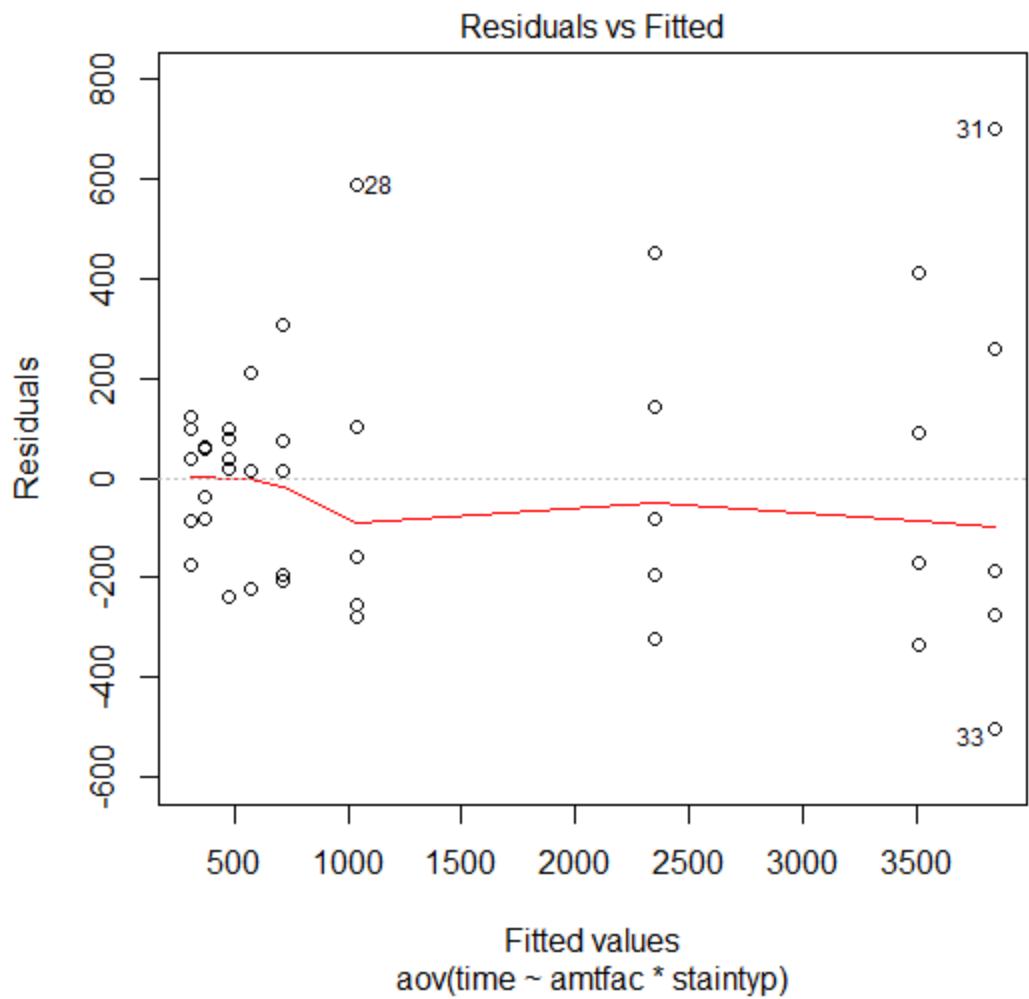
```
> lsmip(m1, staintyp~amtfac)
```



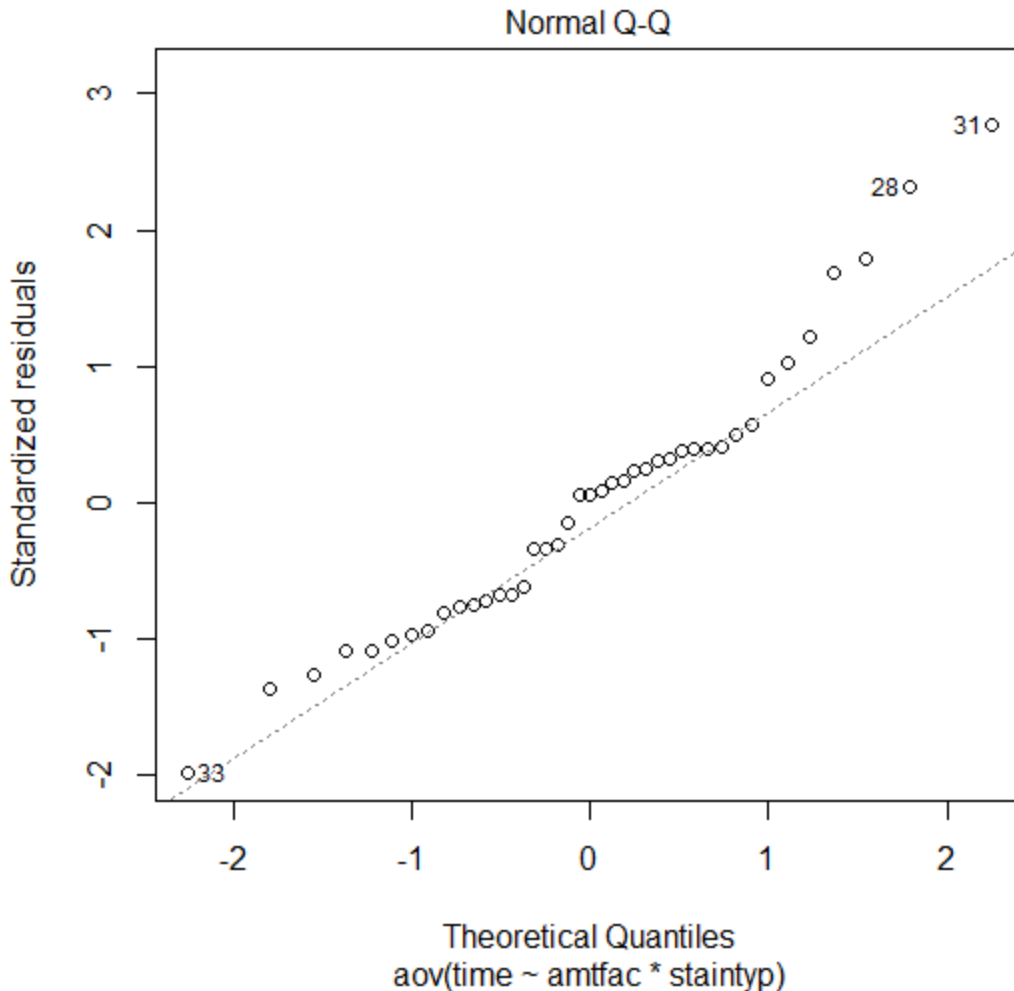
```

>
> #####PARTS 3 & 4
>
> # check resids versus fitted plot and normal QQ plot of resids
> dev.off(which=dev.cur())
null device
      1
> plot(m1,which=1)

```



```
> plot(m1, which=2)
```



```

>
>
> # looks like non-constant variance. Use regression method to estimate
> # transformation:
>
> # to get summary statistics (mean, sd) by treatment we can use the
> # summaryBy function in the doBy package
>
> (sumStats <- summaryBy(time ~ amtfac + staintyp, data = bleach,
+ FUN = function(x) { c(m = mean(x), s = sd(x)) } ) )
  amtfac staintyp   time.m   time.s
1      3      ink 3508.2500 325.91244
2      3      jam  474.8000 137.04452
3      3    tomato  717.4000 212.85277
4      5      ink 2350.8000 305.93659
5      5      jam  370.7500  71.07449
6      5    tomato 1036.4000 361.91339
7      7      ink 3844.2000 479.84550
8      7      jam  312.4000 126.31825
9      7    tomato  570.6667 217.30700
> # produces time.m and time.s for each
> # combination of the levels of amtfac and staintyp
>
> #regress log(SD) on log(mean) to estimate power transformation

```



```
> lm(I(log(time.s)) ~ I(log(time.m)),data=sumStats)
```

```
Call:
```

```
lm(formula = I(log(time.s)) ~ I(log(time.m)), data = sumStats)
```

```
Coefficients:
```

```
(Intercept)  I(log(time.m))  
    1.6941      0.5349
```

```
>
```

```
> # slope is .53 which is approximately 0.5, so this suggests
```

```
> # taking y to the 1-.5=0.5 power (sqrt(y))
```

```
>
```

```
> # creat sqrttime=sqrt(time) in data bleach
```

```
> bleach$sqrttime<-sqrt(bleach$tim)
```

```
>
```

```
> # fit Model 2: 2-way ANOVA model after taking sqrt transformation for time
```

```
> m2 <- aov(sqrttime~amtfac*staintyp,data=bleach)
```

```
> summary(m2)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
amtfac	2	29	14	1.009	0.376
staintyp	2	10446	5223	364.473	< 2e-16 ***
amtfac:staintyp	4	640	160	11.157	8.83e-06 ***
Residuals	32	459	14		

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
> # Type III anova table:
```

```
> Anova(m2,type=3)
```

```
Anova Table (Type III tests)
```

```
Response: sqrttime
```

	Sum Sq	Df	F value	Pr(>F)
(Intercept)	47125	1	3288.4074	< 2.2e-16 ***
amtfac	48	2	1.6702	0.2042
staintyp	10504	2	366.5018	< 2.2e-16 ***
amtfac:staintyp	640	4	11.1566	8.83e-06 ***
Residuals	459	32		

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
>
```

```
> # Now get the marginal and joint lsmeans:
```

```
> lsmeans(m2, specs= list(~staintyp,~amtfac,~amtfac:staintyp))
```

```
$`staintyp` lsmeans`
```

staintyp	lsmean	SE	df	lower.CL	upper.CL
ink	56.49836	1.017349	32	54.42609	58.57063
jam	19.36178	1.017349	32	17.28951	21.43405
tomato	27.32340	1.080598	32	25.12229	29.52450

```
$`amtfac` lsmeans`
```

amtfac	lsmean	SE	df	lower.CL	upper.CL
3	35.76513	1.017349	32	33.69285	37.83740
5	33.14307	1.017349	32	31.07080	35.21534
7	34.27534	1.080598	32	32.07423	36.47644

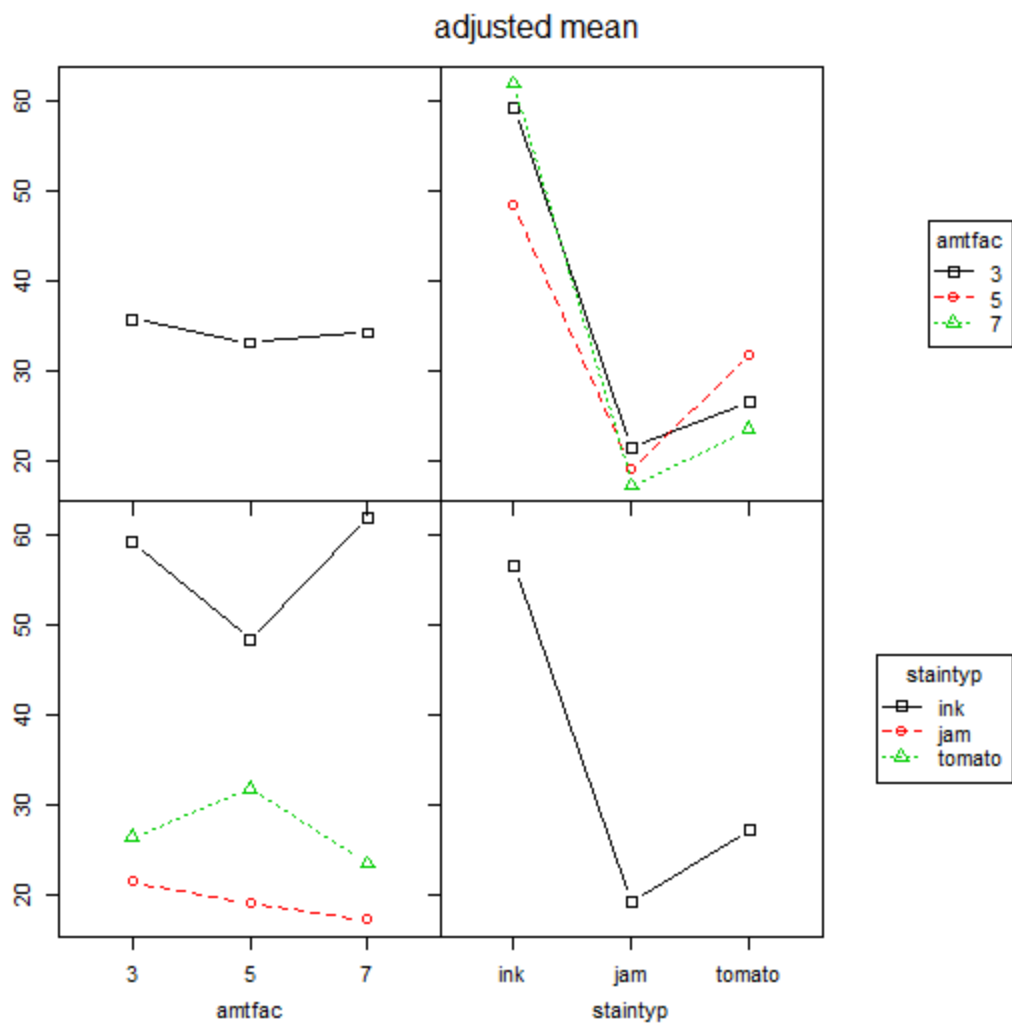
```
$`amtfac:staintyp` lsmeans`
```

amtfac	staintyp	lsmean	SE	df	lower.CL	upper.CL
3	ink	59.18301	1.892799	32	55.32750	63.03851

```
5      ink 48.40488 1.692971 32 44.95641 51.85335
7      ink 61.90719 1.692971 32 58.45872 65.35566
3      jam 21.56007 1.692971 32 18.11160 25.00854
5      jam 19.18658 1.892799 32 15.33108 23.04209
7      jam 17.33869 1.692971 32 13.89022 20.78716
3  tomato 26.55230 1.692971 32 23.10383 30.00077
5  tomato 31.83775 1.692971 32 28.38929 35.28622
7  tomato 23.58014 2.185616 32 19.12818 28.03209
```

Warning messages:

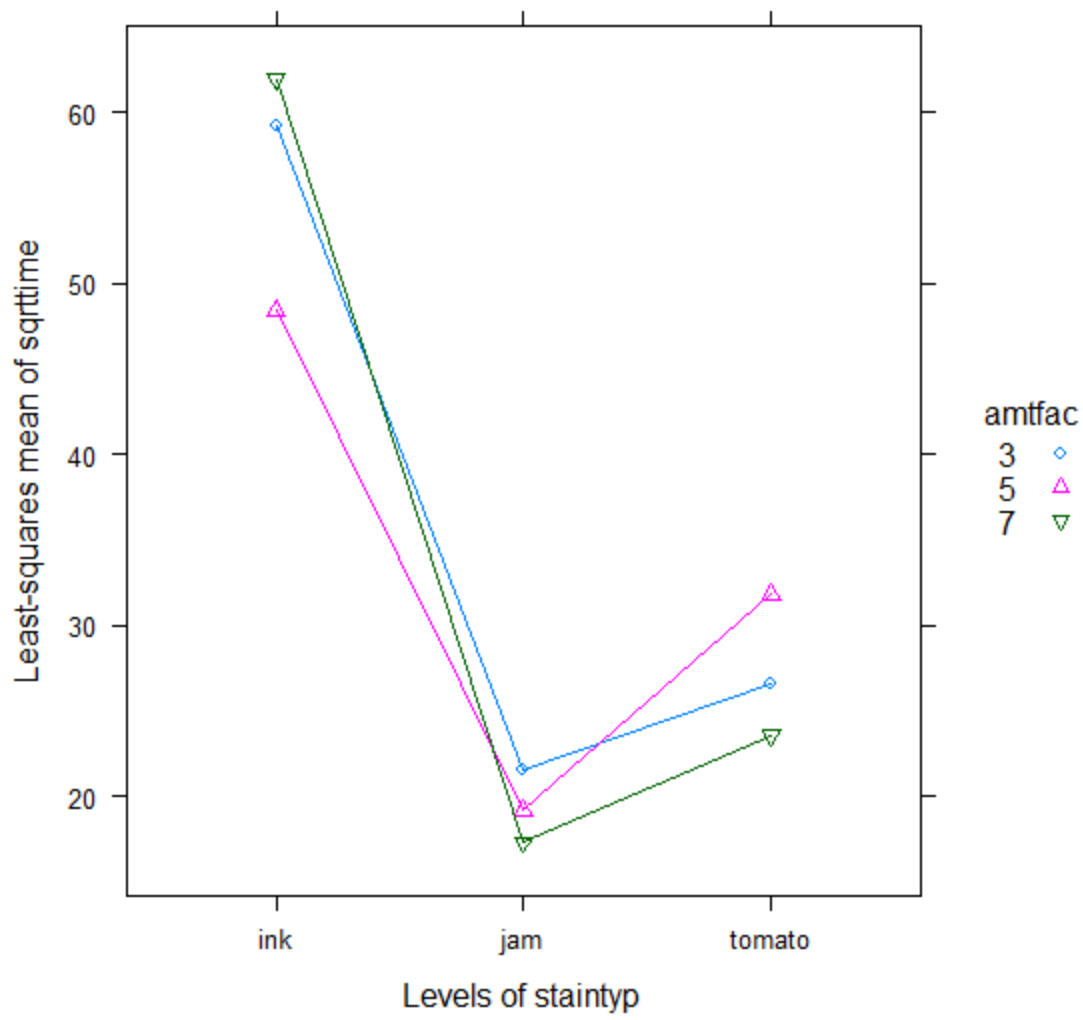
```
1: In lsmeans(m2, specs = list(~staintyp, ~amtfac, ~amtfac:staintyp)) :
  lsmeans of staintyp may be misleading due to interaction with other
predictor(s)
2: In lsmeans(m2, specs = list(~staintyp, ~amtfac, ~amtfac:staintyp)) :
  lsmeans of amtfac may be misleading due to interaction with other predictor(s)
>
> # get profile plots (uses package phia)
> ( trtmns2 <- interactionMeans(m2) )
  amtfac staintyp adjusted mean
1      3      ink      59.18301
2      5      ink      48.40488
3      7      ink      61.90719
4      3      jam      21.56007
5      5      jam      19.18658
6      7      jam      17.33869
7      3  tomato      26.55230
8      5  tomato      31.83775
9      7  tomato      23.58014
> plot(trtmns2)
```



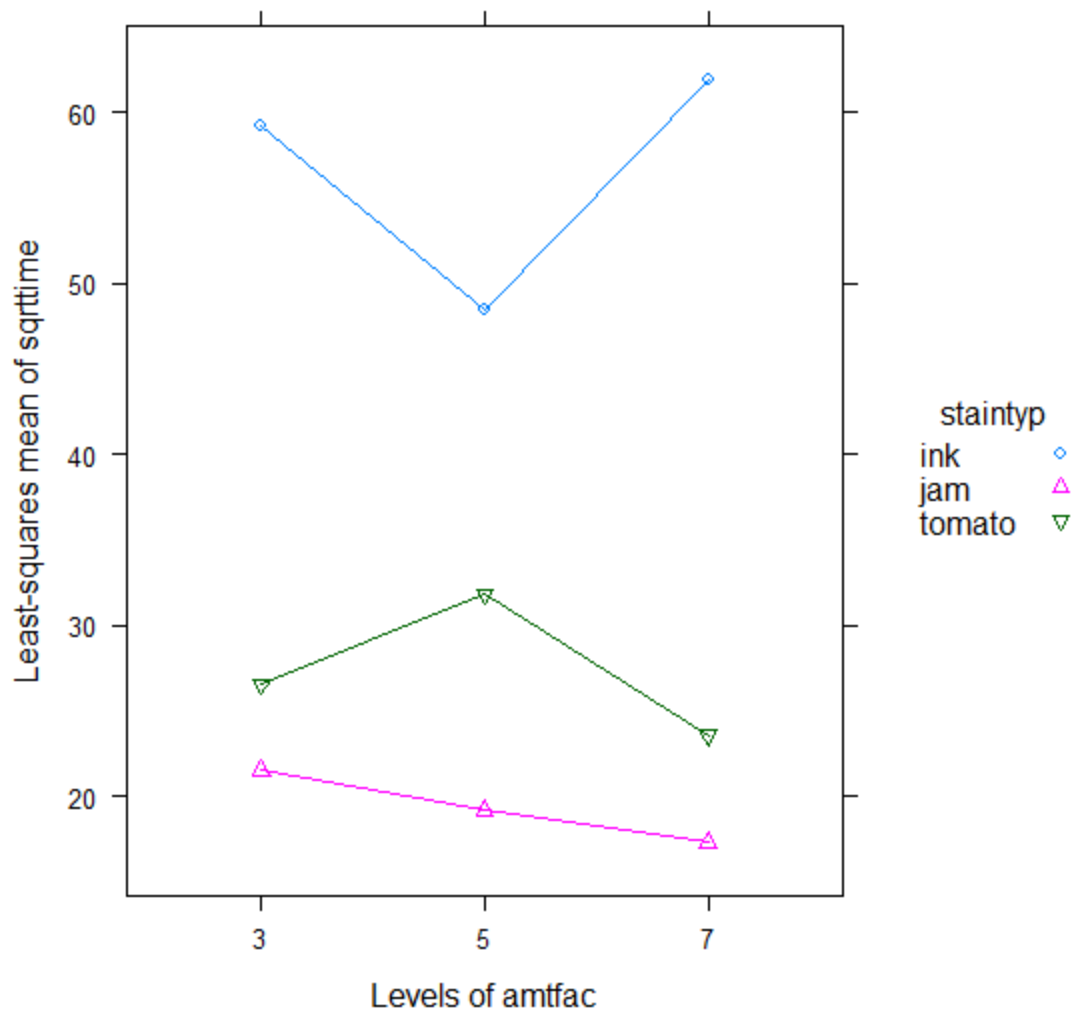
```

>
> # alternatively can use function lsmp from lsmeans package
>
> lsmp(m2, amtfac~staintyp)

```



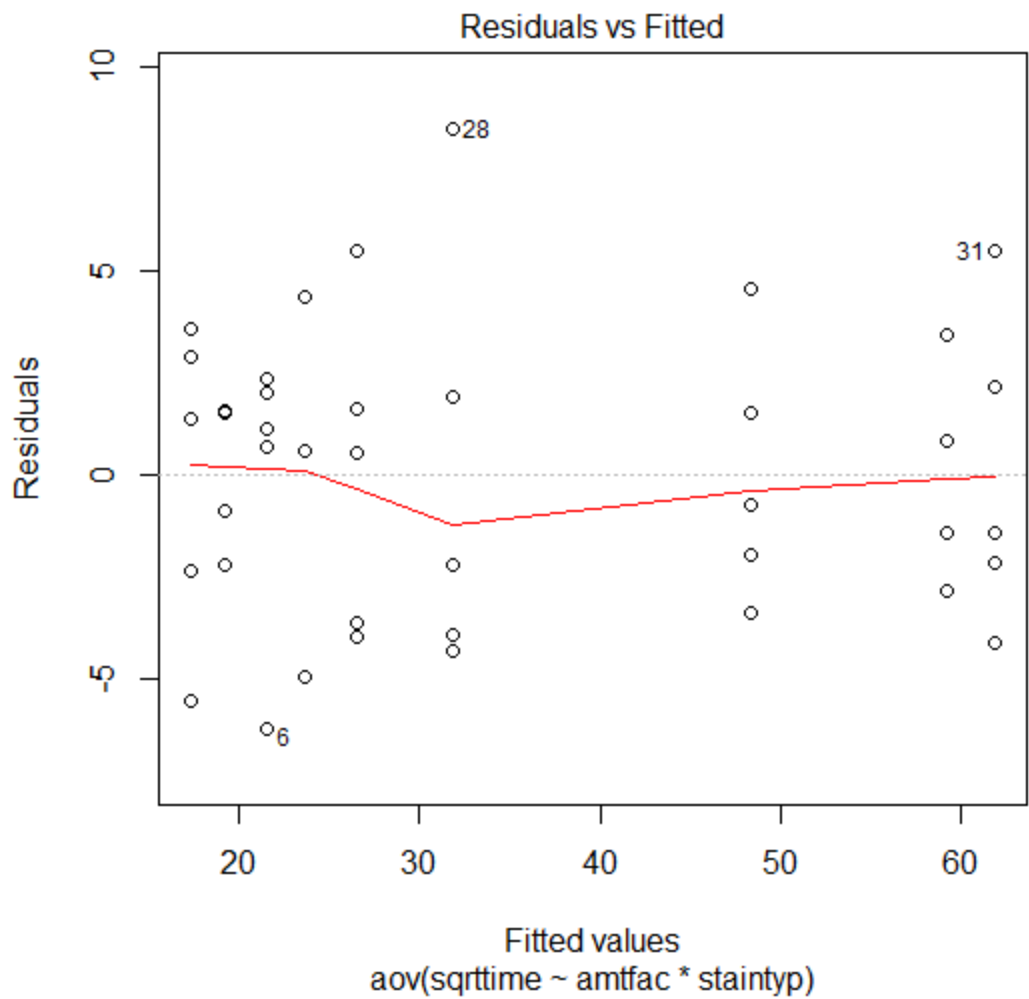
```
> lsmip(m2, staintyp~amtfac)
```



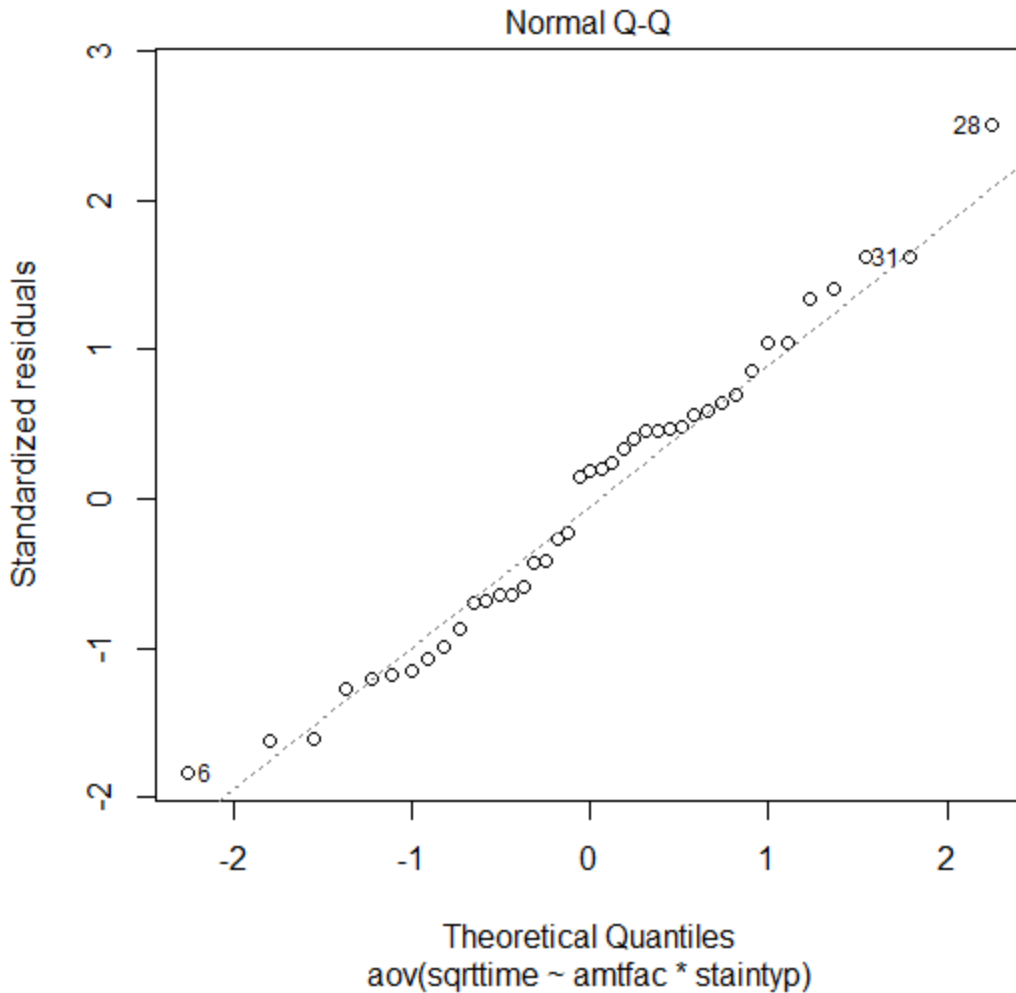
```

>
> # check resid versus fitted plot and normal QQ plot of resid
> dev.off(which=dev.cur())
null device
      1
> plot(m2,which=1)

```



```
> plot(m2, which=2)
```



```

>
> # These look better.
>
> #####PART 5
>
> # below are the linear and quadratic contrasts for amt of bleach within each
> # stain type:
> c1<-c(-1,0,1,0,0,0,0,0,0)
> c2<-c(1,-2,1,0,0,0,0,0,0)
> c3<-c(0,0,0,-1,0,1,0,0,0)
> c4<-c(0,0,0,1,-2,1,0,0,0)
> c5<-c(0,0,0,0,0,0,-1,0,1)
> c6<-c(0,0,0,0,0,0,-1,2,1)
>
> lsmeans(m2, specs=ls~amtfac:staintyp,
+   contr=list(lsm=list(linear.bleach.stain1=c1,
+     nonlinear.bleach.stain1=c2,
+     linear.bleach.stain2=c3,
+     nonlinear.bleach.stain2=c4,
+     linear.bleach.stain3=c5,
+     nonlinear.bleach.stain3=c6)))
$`amtfac:staintyp` lsmeans`
  amtfac staintyp  lsmean      SE df lower.CL upper.CL
    3      ink 59.18301 1.892799 32 55.32750 63.03851

```

```
5      ink 48.40488 1.692971 32 44.95641 51.85335
7      ink 61.90719 1.692971 32 58.45872 65.35566
3      jam 21.56007 1.692971 32 18.11160 25.00854
5      jam 19.18658 1.892799 32 15.33108 23.04209
7      jam 17.33869 1.692971 32 13.89022 20.78716
3  tomato 26.55230 1.692971 32 23.10383 30.00077
5  tomato 31.83775 1.692971 32 28.38929 35.28622
7  tomato 23.58014 2.185616 32 19.12818 28.03209
```

```
$`amtfac:staintyp lsm`
```

	estimate	SE	df	t.ratio	p.value
linear.bleach.stain1	2.7241867	2.539457	32	1.07274	0.29141
nonlinear.bleach.stain1	24.2804466	4.232428	32	5.73677	0.00000
linear.bleach.stain2	-4.2213813	2.394223	32	-1.76315	0.08742
nonlinear.bleach.stain2	0.5255913	4.479181	32	0.11734	0.90732
linear.bleach.stain3	-2.9721692	2.764610	32	-1.07508	0.29038
nonlinear.bleach.stain3	60.7033400	4.371233	32	13.88701	0.00000

```
  p values are not adjusted
```

```
>
> options(op) # reset contrasts to their original values
>
>
```