

## Output from gasadd1.R

```
> library(nlme) # In R there are user-contributed "packages" of function
> # that do extend the basic capabilities of R with all sorts of
> # different statistical methods. Some packages are built into the
> # basic distribution of R (nlme is one such package), others
> # have to be downloaded and installed. Either way, the functions
> # and capabilities in a given package are not available in an R
> # session until the package has been loaded with the command
> # library(package_name).
> # The nlme package is a big one with lots of tools for
> # exploring grouped or clustered data, and for fitting
> # linear and nonlinear mixed effect models. For this example,
> # we are just fitting a simple one-way analysis of variance
> # model, which is a linear model (not a mixed model at all)
> # but there are couple of capabilities in the nlme package
> # that are still convenient and useful here.
>
> # other useful packages used in this script:
> library(MASS)
> library(car)
> library(lsmeans)
>
> gasdata <- read.table(file="gasadd.dat",header=T) # read the data from an
> # external file and grab the
> # variable names from the
> # 1st line of that file.
> # The data are placed into
> # a "data frame", which I've
> # called gasdata
> head(gasdata) #look at the first few rows of gasdata
  add octane
1    0  91.7
2    0  91.2
3    0  90.9
4    0  90.6
5    1  91.7
6    1  91.9
>
> gasdata$addfac <- factor(gasdata$add, ordered=T)
> # by default a numeric variable is not treated when
> # read in, so make a copy of the add variable that
> # R will now treat as a factor, with levels that
> # are assumed to be ordered (rather than qualitative)
>
>
> attach(gasdata) # this makes the variables within gasdata available
The following object(s) are masked from 'gasdata (position 3)':

  add, addfac, octane
The following object(s) are masked from 'gasdata (position 4)':

  add, addfac, octane
The following object(s) are masked from 'gasdata (position 5)':

  add, addfac, octane
```

The following object(s) are masked from 'gasdata (position 10)':

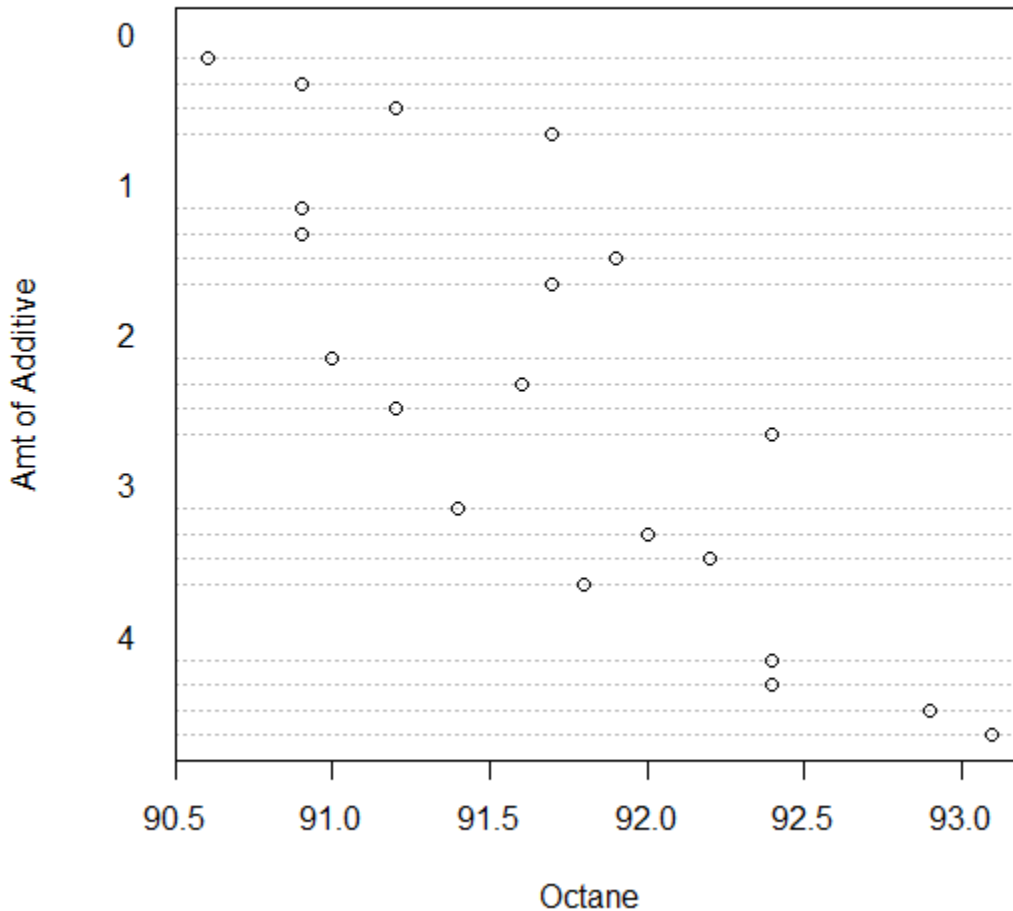
```
add, addfac, octane
```

The following object(s) are masked from 'gasdata (position 11)':

```
add, octane
```

```
> # so you can refer to them as add or octane rather than
> # gasadd$add or gasadd$octane
>
> dotchart(octane,groups=addfac) # plot the octane values with a dotplot
> title(main="Dotplot of Gasoline Additive Data",xlab="Octane",
+   ylab="Amt of Additive") # titles and x and y axis labels can be added
> # with the title() function or the main, xlab
> # and ylab arguments can be added to the original
> # dotchart() function call
```

### Dotplot of Gasoline Additive Data

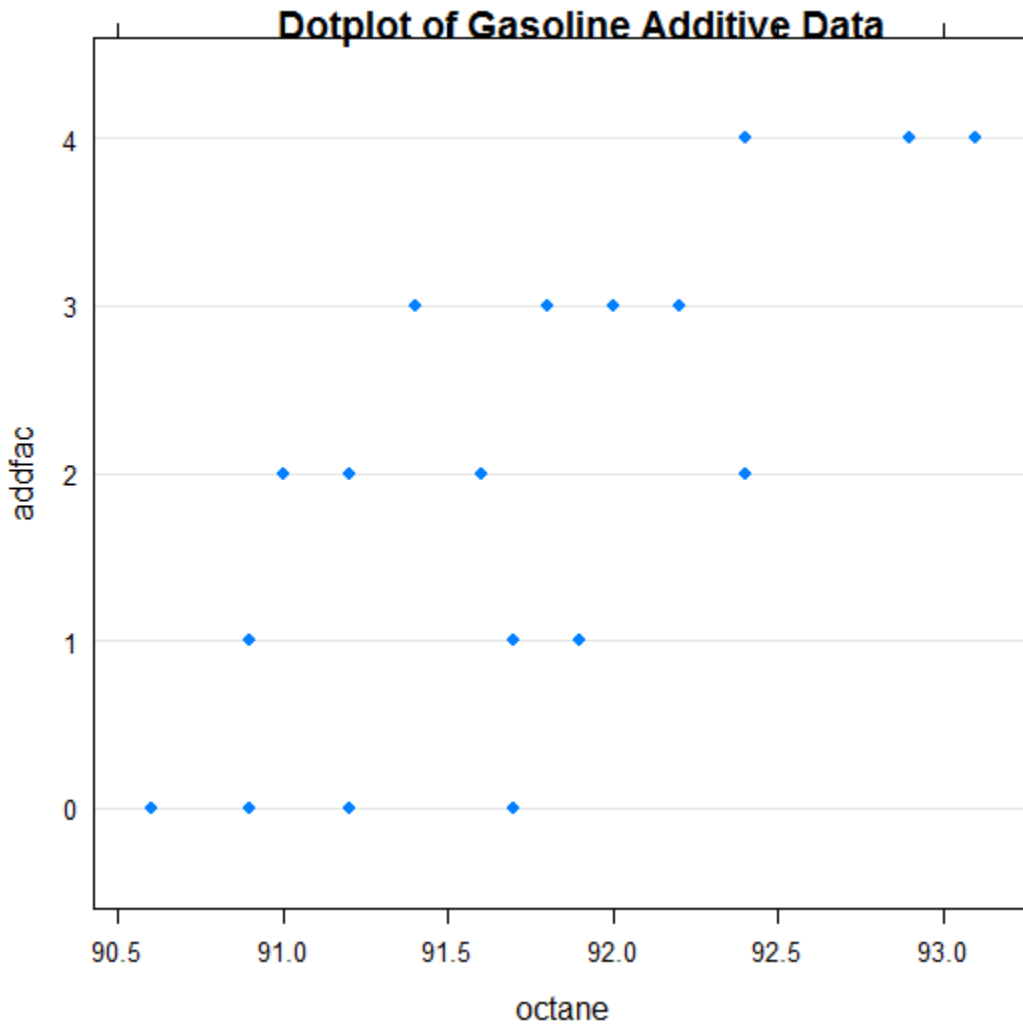


```
>
> # I like the dotplot produced by default when you plot the data after they
> # have been turned into a groupedData object. Such an object is a data frame
> # but with extra information attached to it that comes in the form of a
> # formula that identifies the response variable, (octane below), the
> # covariate or explanatory variable that we want to regard as the "primary"
> # covariate (for plotting and modeling purposes) and any grouping structure.
> # Below the formula is "octane~1|addfac". Here the "~" plays the role of an
> # equal sign, octane is the response (or y variable), 1 represents an
```

```

> # intercept or constant term, and what comes after the vertical bar is
> # the factor that identifies groups.
>
> gasdata2 <- groupedData(octane~1|addfac, gasdata)
>
> plot(gasdata2) # R is an "object-oriented" computing language. That means that
> # quantities that we specify are treated as one of many
> # different types of objects, and functions that operate
> # on those quantities figure out what to do based upon what
> # kind of object that quantity is. E.g., the plot function
> # figure out how to plot its argument based upon the argument's
> # object type. In this case, gasdata2 is a groupedData
> # object, and the default way to handle it is to produce
> # a dotplot.
> title(main="Dotplot of Gasoline Additive Data")

```

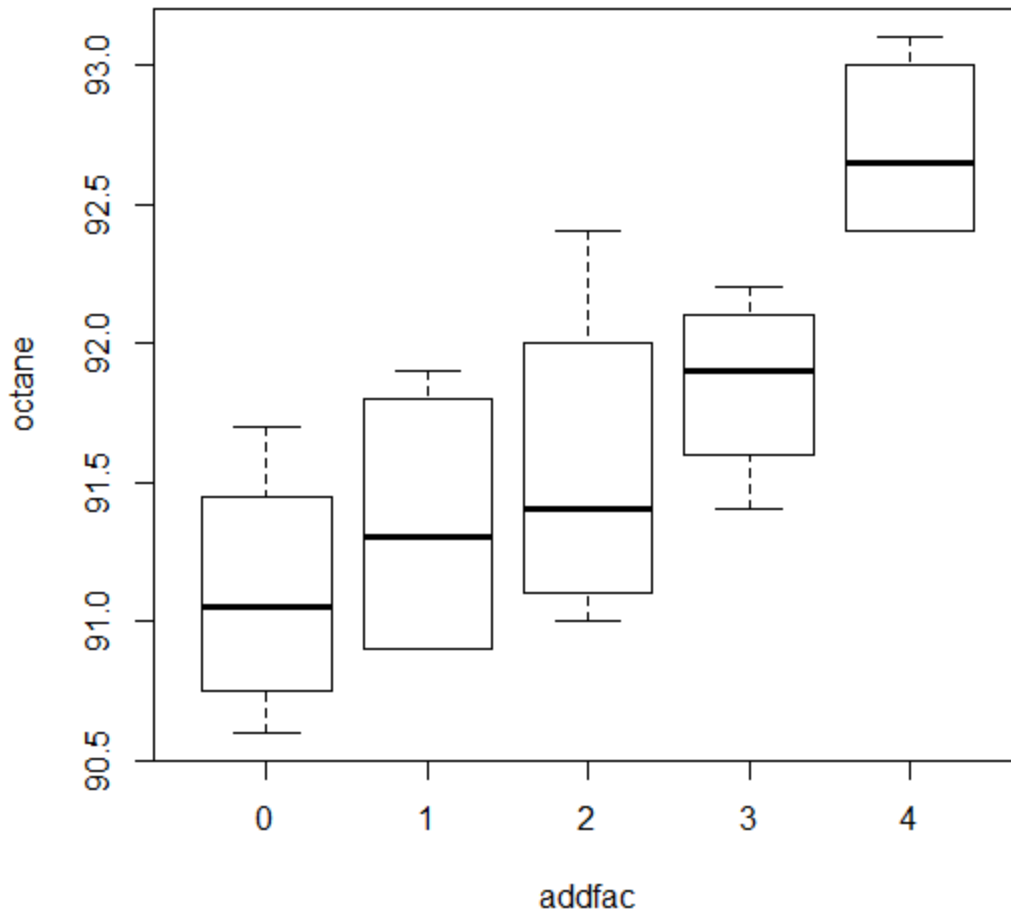


```

>
> # side-by-side boxplots are also useful:
>
> plot(octane~addfac,data=gasdata) # this shows another use of the plot function
> # it can take a formula as argument and
> # figure out what to do
> title(main="Boxplots of octane for each level of additive")

```

## Boxplots of octane for each level of additive



```
>
> # fit a one way anova model using the aov() function. The aov() function
> # is just a wrapper for the lm() function, which fits linear models of all
> # sorts of varieties (regression models, anova models, ancova models).
> # That is, aov() calls lm() where the model fitting is actually done. But
> # this is convenient because the default reporting of a fitted lm is not as
> # suitable for our purposes (not as suitable for an anova model) as the
> # default reporting of a fitted aov model
> m1 <- aov(octane~addfac,gasdata)
> m1      # this "prints" the fitted model
Call:
  aov(formula = octane ~ addfac, data = gasdata)
```

Terms:

	addfac	Residuals
Sum of Squares	6.108	3.370
Deg. of Freedom	4	15

Residual standard error: 0.4739902

Estimated effects are balanced

```
> summary(m1) # this summarizes the fitted model.
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
addfac	4	6.108	1.5270	6.797	0.0025 **
Residuals	15	3.370	0.2247		

```

---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> coefficients(m1) # this gives the actual fitted model parameters
(Intercept)    addfac.L    addfac.Q    addfac.C    addfac^4
91.71000000    1.17004273    0.34743961    0.18973666    0.03585686
>
> summary.lm(m1) # this summarizes the fitted model treating it as generic lm

Call:
aov(formula = octane ~ addfac, data = gasdata)

Residuals:
    Min       1Q   Median       3Q      Max
-0.550 -0.375  0.000   0.350  0.850

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  91.71000    0.10599  865.291 < 2e-16 ***
addfac.L     1.17004    0.23700   4.937 0.000179 ***
addfac.Q     0.34744    0.23700   1.466 0.163291
addfac.C     0.18974    0.23700   0.801 0.435865
addfac^4     0.03586    0.23700   0.151 0.881757
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.474 on 15 degrees of freedom
Multiple R-squared:  0.6444,    Adjusted R-squared:  0.5496
F-statistic: 6.797 on 4 and 15 DF,  p-value: 0.002499

>
> # not necessarily an aov model. This is an easy way to get some
> # additional useful output.
>
>
>
> # now estimate the mean for each level of additive:
> lsmeans(m1, specs = ~ addfac)
$`addfac lsmeans`
  addfac lsmean      SE df lower.CL upper.CL
    0     91.10 0.2369951 15  90.59486  91.60514
    1     91.35 0.2369951 15  90.84486  91.85514
    2     91.55 0.2369951 15  91.04486  92.05514
    3     91.85 0.2369951 15  91.34486  92.35514
    4     92.70 0.2369951 15  92.19486  93.20514

>
>
> # a contrast can be tested using the linearHypothesis function from the
> # car package. We need to define contrasts we are interested and then
> # stack them up as the rows of a matrix which we'll call hypmat1
> # A command like
> # x <- 3
> # assigns 3 to the variable x and produces no output
> # If we put parentheses around that assignment, the assignment is done
> # but in addition, R prints out the value of the variable after the
> # assignment has been executed:
> # (x <- 3)
> # Still assigns 3 to x, but also prints out 3

```

```

>
> ( hypmat1 <- matrix( c(0,1,0,0,0, 0,0,1,0,0, 0,0,0,1,0, 0,0,0,0,1),
+   nrow=4,byrow=T ) )
      [,1] [,2] [,3] [,4] [,5]
[1,]    0    1    0    0    0
[2,]    0    0    1    0    0
[3,]    0    0    0    1    0
[4,]    0    0    0    0    1
>
> linearHypothesis(m1,hypothesis.matrix=hypmat1) # main effect of additive
Linear hypothesis test

Hypothesis:
addfac.L = 0
addfac.Q = 0
addfac.C = 0
addfac^4 = 0

Model 1: restricted model
Model 2: octane ~ addfac

   Res.Df  RSS Df Sum of Sq    F   Pr(>F)
1      19 9.478
2      15 3.370  4      6.108 6.7967 0.002499 **
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
>
> ( hypmat2 <- hypmat1[2:4,] )
      [,1] [,2] [,3] [,4] [,5]
[1,]    0    0    1    0    0
[2,]    0    0    0    1    0
[3,]    0    0    0    0    1
> linearHypothesis(m1,hypothesis.matrix=hypmat2) # nonlinear effects of add
Linear hypothesis test

Hypothesis:
addfac.Q = 0
addfac.C = 0
addfac^4 = 0

Model 1: restricted model
Model 2: octane ~ addfac

   Res.Df  RSS Df Sum of Sq    F Pr(>F)
1      18 4.002
2      15 3.370  3      0.632 0.9377 0.447
>
> # fit simple linear regression of octane on amount of additive
> m2 <- lm (octane~add,data=gasdata)
> summary(m2)

Call:
lm(formula = octane ~ add, data = gasdata)

Residuals:
   Min       1Q   Median       3Q      Max
-0.7100 -0.3875 -0.0600  0.3825  0.7300

```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	90.97000	0.18262	498.139	< 2e-16	***
add	0.37000	0.07455	4.963	0.000101	***

---

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

Residual standard error: 0.4715 on 18 degrees of freedom

Multiple R-squared: 0.5778, Adjusted R-squared: 0.5543

F-statistic: 24.63 on 1 and 18 DF, p-value: 0.0001006

>

> options(op) #reset contrasts to what they were to start (the default)

>