

# Solutions L8

```
knitr::opts_chunk$set(echo = T, message = F,  
  fig.width = 7, fig.height=3.5)  
options(scipen=999)
```

```
library(dplyr)  
library(rgl)
```

## 1 Concepts

### 1.1 Question 1

In equation 1, the parameter  $\beta_0$  measures the value of  $f(X)$  when each of the indicators is zero (i.e. for the baseline region). The parameter  $\beta_k$  measures the difference in  $f(X)$  between the baseline region and the region corresponding to the  $k$ th indicator.

## 2 Applications

```
#####  
### Import and process data ###  
#####  
  
### Import and clean the air quality data  
data("airquality")  
AQ.raw = na.omit(airquality[,1:4])  
  
### Construct new variables  
AQ = AQ.raw  
AQ$TWcp = with(AQ.raw, Temp * Wind)  
AQ$TWrat = with(AQ.raw, Temp / Wind)  
  
#####  
### Helper Functions ###  
#####  
  
### Create function to compute MSPEs  
get.MSPE = function(Y, Y.hat){  
  return(mean((Y - Y.hat)^2))  
}  
  
### Create function which constructs folds for CV  
### n is the number of observations, K is the number of folds  
get.folds = function(n, K) {
```

```

### Get the appropriate number of fold labels
n.fold = ceiling(n / K) # Number of observations per fold (rounded up)
fold.ids.raw = rep(1:K, times = n.fold)
fold.ids = fold.ids.raw[1:n]

### Shuffle the fold labels
folds.rand = fold.ids[sample.int(n)]

return(folds.rand)
}

```

## 2.1 Question 1

```

### Compute medians
med.wind = median(AQ$Wind)
med.temp = median(AQ$Temp)

```

The medians of wind speed and temperature are 9.7 and 79 respectively.

## 2.2 Question 2

The following is a list of all wind speed and temperature values, along with indicators for both variables which are true iff the corresponding variable exceeds its median.

```

### Split wind and temp at their medians, and construct indicators
data = AQ %>%
  mutate(wind.hilo = Wind > med.wind, temp.hilo = Temp > med.temp)

### Print values of wind, temp and their indicators
print(select(data, Wind, Temp, wind.hilo, temp.hilo))

```

```

##      Wind Temp wind.hilo temp.hilo
## 1      7.4   67     FALSE     FALSE
## 2      8.0   72     FALSE     FALSE
## 3     12.6   74      TRUE     FALSE
## 4     11.5   62      TRUE     FALSE
## 7      8.6   65     FALSE     FALSE
## 8     13.8   59      TRUE     FALSE
## 9     20.1   61      TRUE     FALSE
## 12     9.7   69     FALSE     FALSE
## 13     9.2   66     FALSE     FALSE
## 14    10.9   68      TRUE     FALSE
## 15    13.2   58      TRUE     FALSE
## 16    11.5   64      TRUE     FALSE
## 17    12.0   66      TRUE     FALSE
## 18    18.4   57      TRUE     FALSE
## 19    11.5   68      TRUE     FALSE
## 20     9.7   62     FALSE     FALSE
## 21     9.7   59     FALSE     FALSE
## 22    16.6   73      TRUE     FALSE
## 23     9.7   61     FALSE     FALSE
## 24    12.0   61      TRUE     FALSE

```

## 28	12.0	67	TRUE	FALSE
## 29	14.9	81	TRUE	TRUE
## 30	5.7	79	FALSE	FALSE
## 31	7.4	76	FALSE	FALSE
## 38	9.7	82	FALSE	TRUE
## 40	13.8	90	TRUE	TRUE
## 41	11.5	87	TRUE	TRUE
## 44	8.0	82	FALSE	TRUE
## 47	14.9	77	TRUE	FALSE
## 48	20.7	72	TRUE	FALSE
## 49	9.2	65	FALSE	FALSE
## 50	11.5	73	TRUE	FALSE
## 51	10.3	76	TRUE	FALSE
## 62	4.1	84	FALSE	TRUE
## 63	9.2	85	FALSE	TRUE
## 64	9.2	81	FALSE	TRUE
## 66	4.6	83	FALSE	TRUE
## 67	10.9	83	TRUE	TRUE
## 68	5.1	88	FALSE	TRUE
## 69	6.3	92	FALSE	TRUE
## 70	5.7	92	FALSE	TRUE
## 71	7.4	89	FALSE	TRUE
## 73	14.3	73	TRUE	FALSE
## 74	14.9	81	TRUE	TRUE
## 76	14.3	80	TRUE	TRUE
## 77	6.9	81	FALSE	TRUE
## 78	10.3	82	TRUE	TRUE
## 79	6.3	84	FALSE	TRUE
## 80	5.1	87	FALSE	TRUE
## 81	11.5	85	TRUE	TRUE
## 82	6.9	74	FALSE	FALSE
## 85	8.6	86	FALSE	TRUE
## 86	8.0	85	FALSE	TRUE
## 87	8.6	82	FALSE	TRUE
## 88	12.0	86	TRUE	TRUE
## 89	7.4	88	FALSE	TRUE
## 90	7.4	86	FALSE	TRUE
## 91	7.4	83	FALSE	TRUE
## 92	9.2	81	FALSE	TRUE
## 93	6.9	81	FALSE	TRUE
## 94	13.8	81	TRUE	TRUE
## 95	7.4	82	FALSE	TRUE
## 99	4.0	89	FALSE	TRUE
## 100	10.3	90	TRUE	TRUE
## 101	8.0	90	FALSE	TRUE
## 104	11.5	86	TRUE	TRUE
## 105	11.5	82	TRUE	TRUE
## 106	9.7	80	FALSE	TRUE
## 108	10.3	77	TRUE	FALSE
## 109	6.3	79	FALSE	FALSE
## 110	7.4	76	FALSE	FALSE
## 111	10.9	78	TRUE	FALSE
## 112	10.3	78	TRUE	FALSE
## 113	15.5	77	TRUE	FALSE

```
## 114 14.3 72 TRUE FALSE
## 116 9.7 79 FALSE FALSE
## 117 3.4 81 FALSE TRUE
## 118 8.0 86 FALSE TRUE
## 120 9.7 97 FALSE TRUE
## 121 2.3 94 FALSE TRUE
## 122 6.3 96 FALSE TRUE
## 123 6.3 94 FALSE TRUE
## 124 6.9 91 FALSE TRUE
## 125 5.1 92 FALSE TRUE
## 126 2.8 93 FALSE TRUE
## 127 4.6 93 FALSE TRUE
## 128 7.4 87 FALSE TRUE
## 129 15.5 84 TRUE TRUE
## 130 10.9 80 TRUE TRUE
## 131 10.3 78 TRUE FALSE
## 132 10.9 75 TRUE FALSE
## 133 9.7 73 FALSE FALSE
## 134 14.9 81 TRUE TRUE
## 135 15.5 76 TRUE FALSE
## 136 6.3 77 FALSE FALSE
## 137 10.9 71 TRUE FALSE
## 138 11.5 71 TRUE FALSE
## 139 6.9 78 FALSE FALSE
## 140 13.8 67 TRUE FALSE
## 141 10.3 76 TRUE FALSE
## 142 10.3 68 TRUE FALSE
## 143 8.0 82 FALSE TRUE
## 144 12.6 64 TRUE FALSE
## 145 9.2 71 FALSE FALSE
## 146 10.3 81 TRUE TRUE
## 147 10.3 69 TRUE FALSE
## 148 16.6 63 TRUE FALSE
## 149 6.9 70 FALSE FALSE
## 151 14.3 75 TRUE FALSE
## 152 8.0 76 FALSE FALSE
## 153 11.5 68 TRUE FALSE
```

## 2.3 Question 3

The following is a summary of the linear model of ozone on the indicators for wind speed and temperature.

```
fit.main = lm(Ozone ~ wind.hilo + temp.hilo, data = data)
```

```
summary(fit.main)
```

```
##
## Call:
## lm(formula = Ozone ~ wind.hilo + temp.hilo, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -53.256 -13.065  -1.874   9.435  98.744
##
```

```
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)    36.216      4.547   7.966 0.00000000000018 ***
## wind.hiloTRUE  -21.342      4.973  -4.292 0.0000387938694 ***
## temp.hiloTRUE   33.040      4.970   6.648 0.0000000012523 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 24.82 on 108 degrees of freedom
## Multiple R-squared:  0.4538, Adjusted R-squared:  0.4437
## F-statistic: 44.86 on 2 and 108 DF,  p-value: 0.00000000000000657

### Extract p-values
anova.table = summary(fit.main)$coef
p.vals = anova.table[,4]
p.val.wind = p.vals["wind.hiloTRUE"]
p.val.temp = p.vals["temp.hiloTRUE"]
```

Both predictors have statistically significant effects on ozone (wind speed:  $p = 3.88\text{e-}05$ , temperature:  $p = 1.25\text{e-}09$ ).

```
### Construct a grid in Temp and Wind
vals.wind = seq(from = 2, to = 21, by = 1)
vals.temp = seq(from = 56, to = 98, by = 2)
pred.grid = expand.grid(Wind = vals.wind, Temp = vals.temp)

### Get indicators on grid
pred.grid.ind = mutate(pred.grid, wind.hilo = Wind > med.wind,
  temp.hilo = Temp > med.temp)

### Get fitted values on our grid
pred.ozone = predict(fit.main, newdata = pred.grid.ind)

### Plot fitted surface with scatterplot
open3d()
persp3d(x = vals.wind, y = vals.temp, z = pred.ozone, col = "orange",
  xlab = "Wind Speed", ylab = "Temperature", zlab = "Ozone")
with(data, points3d(Ozone ~ Wind + Temp))
```

See Figure 1 for a plot of the predicted values from our model. We see that ozone increases for large values of temperature, and decreases for large values of wind speed.

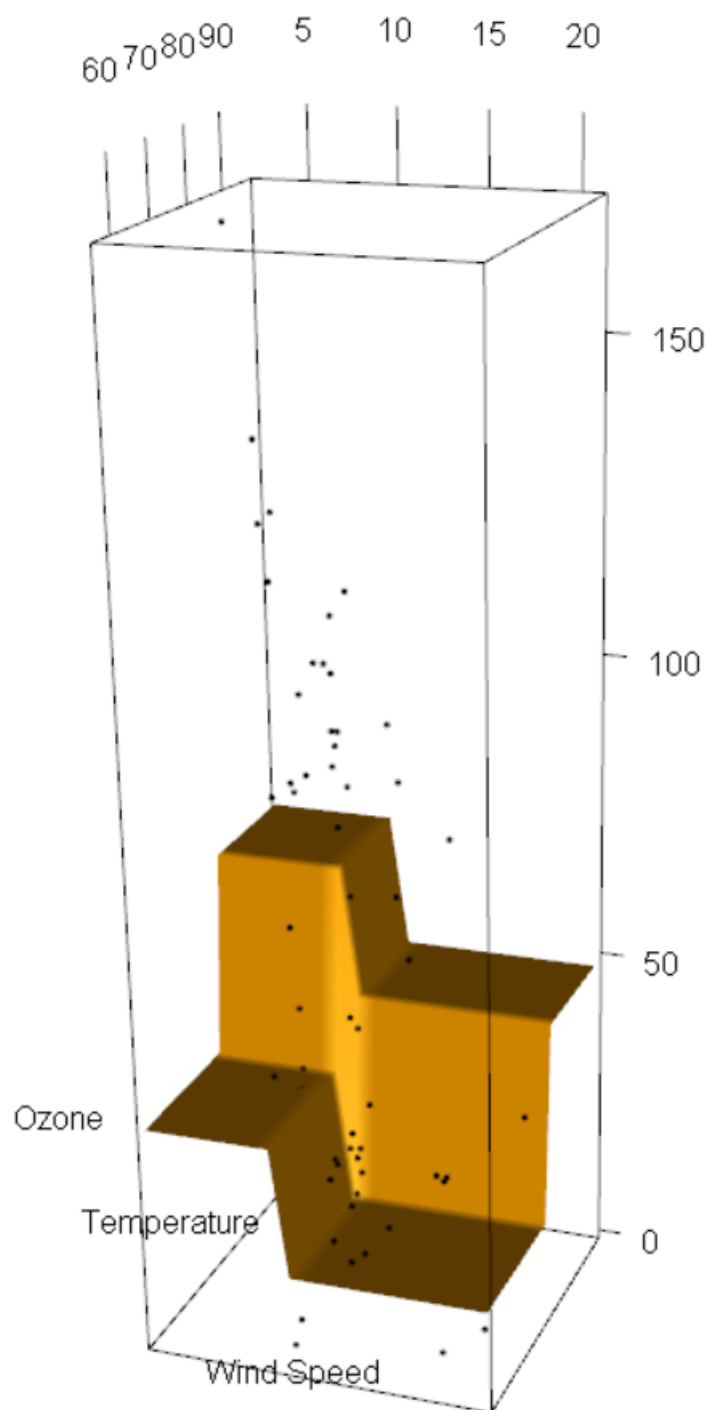
## 2.4 Question 4

The following is a summary of the linear model of ozone on the indicators for wind speed and temperature with interactions.

```
fit.int = lm(Ozone ~ wind.hilo * temp.hilo, data = data)

summary(fit.int)

##
## Call:
## lm(formula = Ozone ~ wind.hilo * temp.hilo, data = data)
##
```



Effects Plot.png

Figure 1: Step function with only main effects.

```
## Residuals:
##      Min       1Q   Median       3Q      Max
## -56.865 -10.220  -0.583   10.135   95.135
##
## Coefficients:
##              Estimate Std. Error t value    Pr(>|t|)
## (Intercept)      29.857      5.322   5.610 0.00000016038 ***
## wind.hiloTRUE     -11.274      6.697  -1.683   0.0952 .
## temp.hiloTRUE      43.008      6.664   6.454 0.00000000326 ***
## wind.hiloTRUE:temp.hiloTRUE -21.532      9.794  -2.198   0.0301 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 24.39 on 107 degrees of freedom
## Multiple R-squared:  0.4774, Adjusted R-squared:  0.4627
## F-statistic: 32.58 on 3 and 107 DF,  p-value: 0.00000000000004851

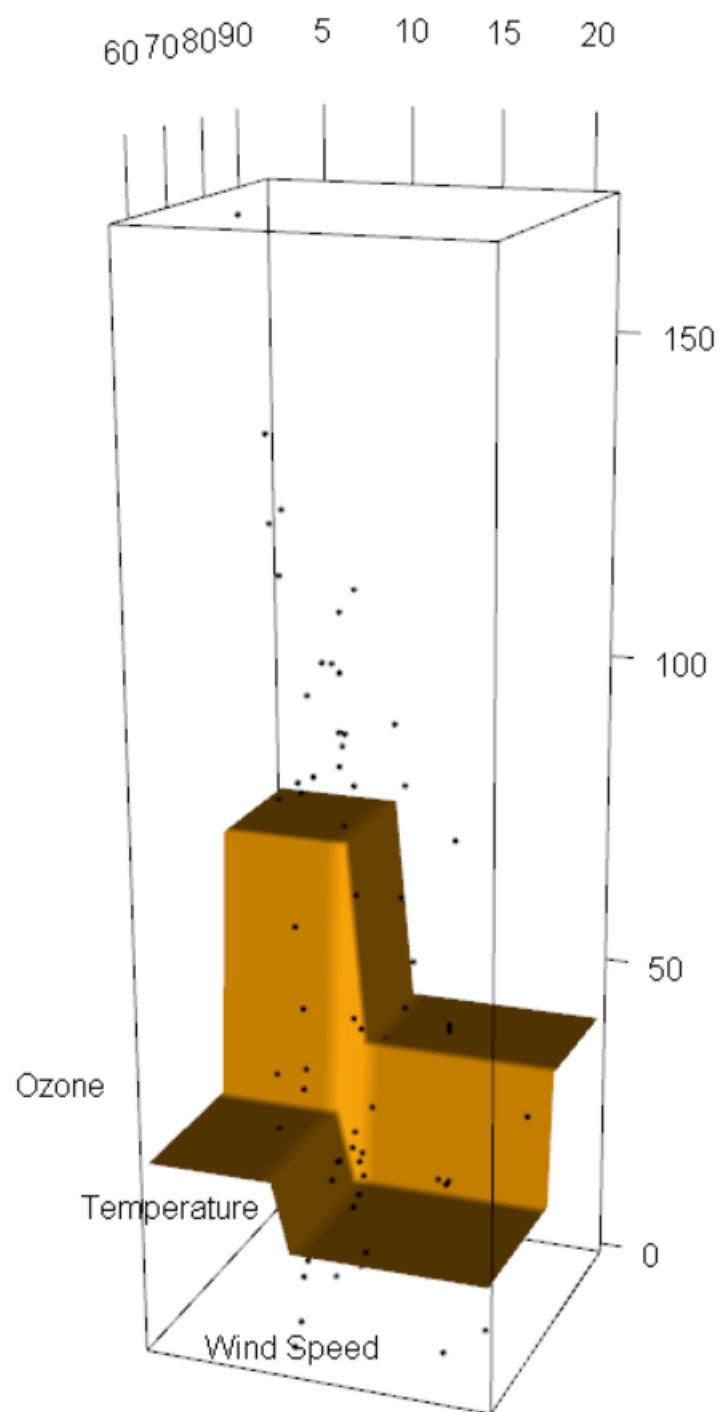
### Extract p-values
anova.table = summary(fit.int)$coef
p.vals = anova.table[,4]
p.val.int = p.vals["wind.hiloTRUE:temp.hiloTRUE"]
```

The interaction term has a statistically significant effect on ozone ( $p = 3.01\text{e-}02$ ).

```
### Get fitted values on our grid
pred.ozone.int = predict(fit.int, newdata = pred.grid.ind)

### Plot fitted surface with scatterplot
open3d()
persp3d(x = vals.wind, y = vals.temp, z = pred.ozone.int, col = "orange",
        xlab = "Wind Speed", ylab = "Temperature", zlab = "Ozone")
with(data, points3d(Ozone ~ Wind + Temp))
```

See Figure 2 for a plot of the predicted values from our model. The interaction appears to lessen the effect of temperature on ozone at large values of wind speed.



Plot.png

Figure 2: Step function with interaction.