

IS609 Week 6 HW

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Page 251, Question 2. Nutritional Requirements

A rancher has determined that the minimum weekly nutritional requirements for an average sized horse include 40lb of protein, 20lb of carbohydrates, and 45lb of roughage. These are obtained from the following sources in varying amounts at the prices indicated:

	Protein (lb)	Carbs (lb)	Roughage (lb)	Cost
Hay (per bale)	0.5	2.0	5.0	\$1.80
Oats (per sack)	1.0	4.0	2.0	\$3.50
Feeding Blocks (per block)	2.0	0.5	1.0	\$0.40
High Protein Conc. (per sack)	6.0	1.0	2.5	\$1.00
Requirements per horse (per week)	40.0	20.0	45.0	

Relationships:

$$\text{Protein} = (0.5 \times \text{Hay}) + (1 \times \text{Oats}) + (2 \times \text{FB}) + (6 \times \text{HPC})$$

$$\text{Carbs} = (92 \times \text{Hay}) + (4 \times \text{Oats}) + (0.5 \times \text{FB}) + (1 \times \text{HPC})$$

$$\text{Roughage} = (5 \times \text{Hay}) + (2 \times \text{Oats}) + (1 \times \text{FB}) + (2.5 \times \text{HPC})$$

$$\text{Cost} = (1.8 \times \text{Hay}) + (3.5 \times \text{Oats}) + (0.4 \times \text{FB}) + (1 \times \text{HPC})$$

Constraints:

$$\text{Protein} > 40$$

$$\text{Carbs} > 20$$

$$\text{Roughage} > 45$$

Optimization:

Minimize Cost

Page 264, Question 6

Minimize $10x + 35y$ subject to:

(Board feet of lumber, hours of carpentry, demand, nonnegativity)

$$8x + 6y \leq 48$$

$$4x + y \leq 20$$

$$y \geq 5$$

$$x, y \geq 0$$

I'll solve all of these for y and graph them out in ggplot, to get an idea of my feasible zone:

```
x <- seq(0,5,by=0.1)

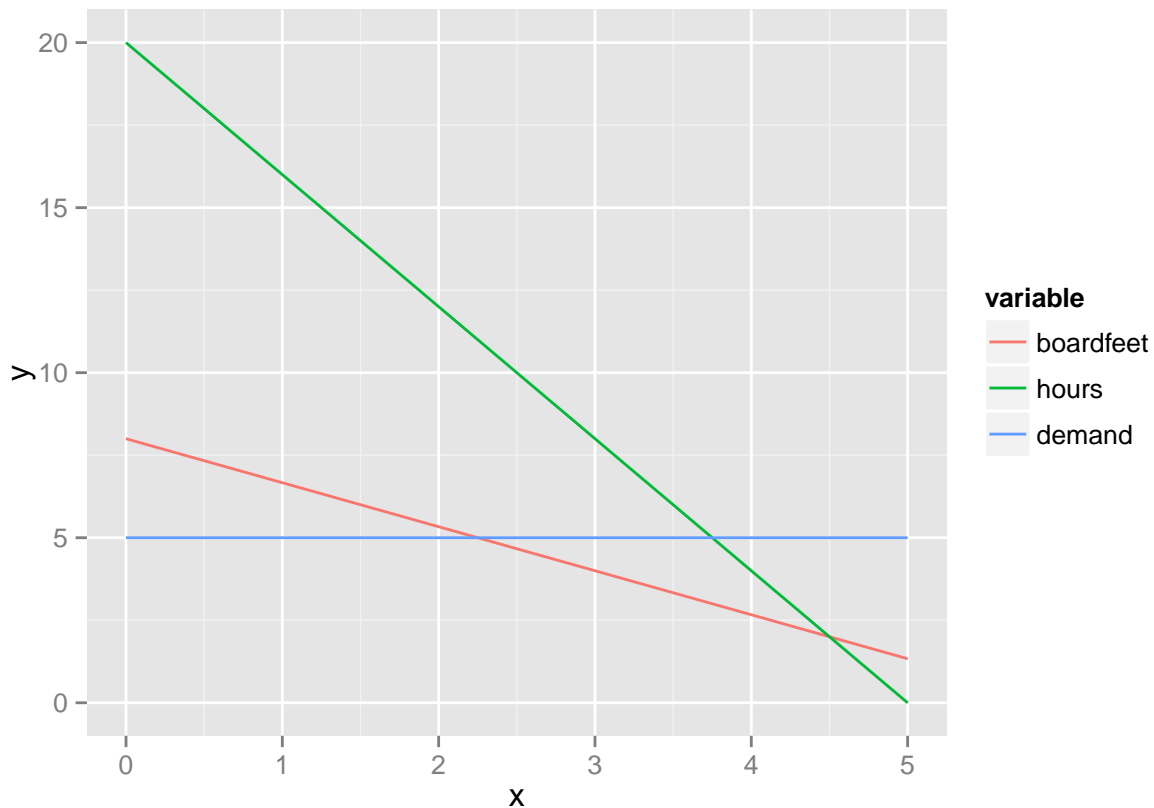
data <- data.frame(x = x, boardfeet = (48-8*x)/6, hours = 20-4*x, demand = 5)

library(reshape2)

data <- melt(data, id.vars="x", value.name="y")

library(ggplot2)

ggplot(data, aes(x=x, y=y, color=variable)) + geom_line()
```



My feasible zone is below the red line (board-feet of lumber), below the green line (hours of carpentry), above the blue line (demand), and above 0. So, the feasible zone is the triangle defined by $y=5$, $x=0$, and the boardfeet line $8x + 6y = 48$. This means that the slope of the objective function will determine which point, (0,8) or (2.25,5) the optimization occurs. If the slope is more negative than the boardfeet line ($-\frac{4}{3}$), then the optimized point will occur at (2.25, 5). Otherwise, it will occur at (0,8).

The slope of the line $10x + 35y = c$ is $-\frac{10}{35}$, or $-\frac{2}{7}$. This is not more negative than the boardfeet slope, so the optimum will occur at (0,8). Lets graph this out.

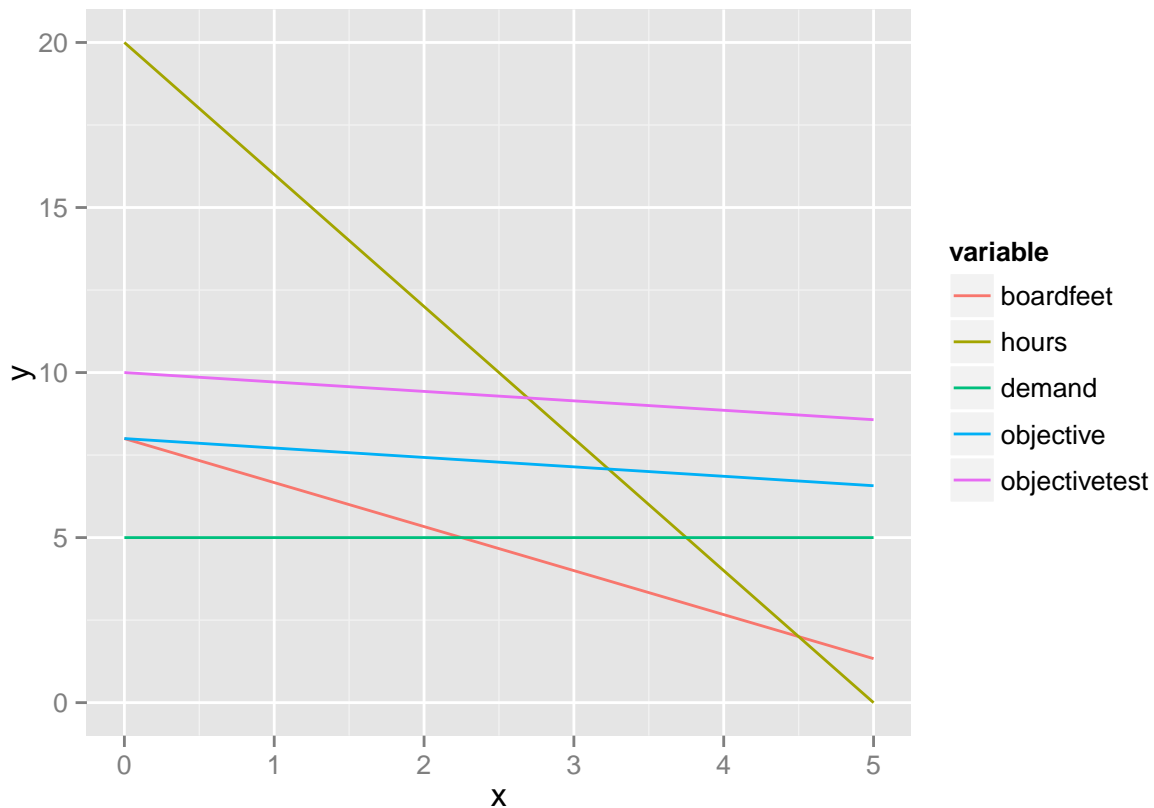
If the line $10x + 35y = c$ crosses through $(0,8)$, then $35 \cdot 8 = c$, so $c = 280$. Lets add this to our lines:

```
x <- seq(0,5,by=0.1)

data <- data.frame(x = x, boardfeet = (48-8*x)/6, hours = 20-4*x, demand = 5,
                  objective = (280-10*x)/35, objectivetest = (350-10*x)/35)

data <- melt(data, id.vars="x", value.name="y")

ggplot(data, aes(x=x, y=y, color=variable)) + geom_line()
```



Here I graphed the objective along with a test, with c a bit higher (at 350). You can see if it's a bit higher, it's outside of the feasible zone.

Page 268, Question 6 (Algebraic Solution)

This time we're going to rewrite the equations in question 6

Maximize $10x + 35y$ subject to:

$$8x + 6y + c1 = 48$$

$$4x + y + c2 = 20$$

$$y + c3 = 5$$

Here we won't consider the case of setting $y = 0$, because we already know $y \geq 5$. So, we'll be solving this by setting four variables to 0 two at a time: x , $c1$, $c2$, and $c3$.

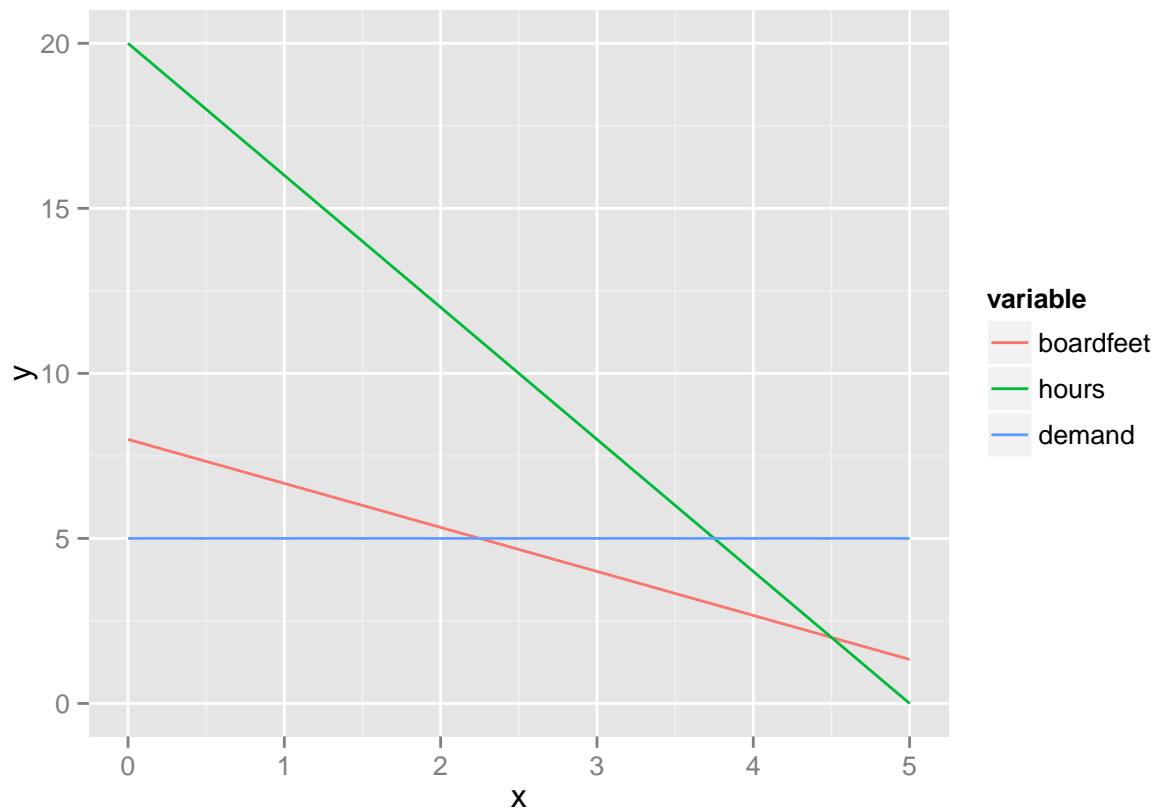
Lets look at the graph again:

```
x <- seq(0,5,by=0.1)

data <- data.frame(x = x, boardfeet = (48-8*x)/6, hours = 20-4*x, demand = 5)

data <- melt(data, id.vars="x", value.name="y")

ggplot(data, aes(x=x, y=y, color=variable)) + geom_line()
```



$c1, c2 = 0$

$$8x + 6y = 48$$

$$4x + y = 20$$

$$y - c3 = 5$$

This corresponds to the point $(4.5, 2)$, with $c3 = -3$, $x = 4.5$, and $y = 2$. This is not feasible because $c3$ is negative. This corresponds to the intersection of the red and green lines.

$c1, c3 = 0$

$$8x + 6y = 48$$

$$4x + y + c2 = 20$$

$$y = 5$$

This corresponds to the point (2.25, 5), with $c2 = 6$, $x=2.25$, and $y = 5$. This is a feasible point, and on the graph it corresponds to the intersection of the blue and red lines.

$$c2, c3 = 0$$

$$8x + 6y + c1 = 48$$

$$4x + y = 20$$

$$y = 5$$

This corresponds with the values $c1 = -12$, $x = 3.75$, and $y = 5$. We know this isn't a feasible point because $c1$ is negative, and the three slack variables we picked need to be nonnegative. This point is (3.75, 5), and corresponds to the intersection of the green and blue lines.

$$c1, x = 0$$

$$6y = 48$$

$$y + c2 = 20$$

$$y - c3 = 5$$

This corresponds with the values $c2 = 12$, $c3 = 3$, and $y = 8$. This is feasible, and corresponds with the point (0,8), which is feasible. This is at the y-intercept of the red line.

$$c2, x = 0$$

$$6y + c1 = 48$$

$$y = 20$$

$$y - c3 = 5$$

This corresponds with the values $c1 = -71$, $c3 = 15$, and $y = 20$. This is not feasible since $c1$ is negative, and corresponds to the point (0,20). This is the y-intercept of the green line.

$$c3, x = 0$$

$$6y + c1 = 48$$

$$y + c2 = 20$$

$$y = 5$$

This corresponds with the values $c1 = 18$, $c2 = 15$, and $y = 5$. This corresponds to the point (0,5), and is feasible. This corresponds with the y-intercept of the blue line.

Solving algebraically, we have three feasible points: (0,5), (0,8), and (2.25, 5). Lets plug these back into the objective function:

x	y	10x + 35y
0	5	175
0	8	280
2.25	5	197.5

This gives us the same answer we had before, the point that maximizes this function is at (0,8).

Page 278, Question 6 (Simplex Solution)

See the write-ups below:

X	Y	C_1	C_2	C_3	Z	RHS	Ratio
8	6	1	6	0	0	48	$48/6 = 8$
4	1	0	1	0	0	20	$20/1 = 20$
0	-1	0	0	1	0	-5	$-5/1 = \textcircled{5}$
-10	$\textcircled{-35}$	0	0	0	1	0	

Dependent variables: $\{C_1, C_2, C_3, Z\}$
 Independent variables: $\{X=0, Y=0\}$
 extreme point: $(X, Y) = (0, 5)$
 Value of objective: $Z = 175$

Optimality test: entering variable is Y
 Feasibility test: Ratio suggests that C_3 is the exiting variable
 Pivot: divide the row containing the exiting variable by the coefficient of the entering variable in this row. Then, eliminate the entering variable Y from the remaining rows

$$Y - C_3 = 5, \quad Y = C_3 + 5$$

$8x + 6C_3 + 30 + C_1 = 48$	$8x + C_1 + 6C_3 = 18$
$4x + C_3 + 5 + C_2 = 20$	$4x + C_2 + C_3 = 15$
$-10x - 35C_3 - 175 + Z = 0$	$-10x - 35C_3 + Z = 175$

X	Y	C_1	C_2	C_3	Z	RHS	Ratio
8	0	1	0	6	0	18	$18/6 = \textcircled{3}$
4	0	0	1	1	0	15	$15/1 = 15$
0	1	0	0	-1	0	5	$5/-1 = -5$
-10	0	0	0	$\textcircled{-35}$	1	175	

Dependent variables: $\{Y, C_1, C_2, Z\}$
 Independent variables: $\{C_3=0, X=0\}$
 optimality Test: entering variable is C_3
 Feasibility Test: exiting variable is C_1
 Pivot: divide the row containing the exiting variable by the coefficient of the entering variable, Then, eliminate the entering variable in all remaining rows

$$\begin{array}{l|l}
\frac{4}{3}x + \frac{1}{6}C_1 + C_3 = 3 & C_3 = 3 - \frac{4}{3}x - \frac{1}{6}C_1 \\
4x + C_2 + 3 - \frac{4}{3}x - \frac{1}{6}C_1 = 15 & \frac{8}{3}x - \frac{1}{6}C_1 + C_2 = 12 \\
y - 3 + \frac{4}{3}x + \frac{1}{6}C_1 = 5 & \frac{4}{3}x + y + \frac{1}{6}C_1 = 8 \\
-10x - 105 + \frac{140}{3}x + \frac{35}{6}C_1 + 2 = 175 & \frac{110}{3}x + \frac{35}{6}C_1 + 2 = 280
\end{array}$$

X	Y	C ₁	C ₂	C ₃	Z	RHS
$\frac{4}{3}$	0	$\frac{1}{6}$	0	1	0	3
$\frac{8}{3}$	0	$-\frac{1}{6}$	1	0	0	12
$\frac{4}{3}$	1	$\frac{1}{6}$	0	0	0	8
$\frac{110}{3}$	0	$\frac{35}{6}$	0	0	1	280

Dependent Variables: $\{Y, C_2, C_3, Z\}$
Independent Variables: $\{X=0, C_1=0\}$

There are no more negative coefficients in the Tableau, so
This is our solution. Let's plug in $C_1=0$ to get our y

$$\begin{aligned}
8x + 6y + C_1 &= 48 \quad x=0, C_1=0 \\
6y &= 48 \\
y &= 8
\end{aligned}$$

So, we have our point: $(x=0, y=8)$
and the profit at this point: \$280

Page 284, Question 1

For the example problem in this section, determine the sensitivity of the optimal solution to a change in c_2 using the objective function $25x_1 + c_2x_2$

In the example program we can build on the fact that we know that the extreme point (12,15) remains optimal as long as the slope of the objective function is less than $-\frac{2}{3}$ but greater than $\frac{5}{4}$. With c_2 variable, the slope is $-\frac{25}{c_2}$. So, we end up with the inequality:

$$-\frac{5}{4} \leq -\frac{25}{c_2} \leq -\frac{2}{3}$$

Or

$$20 \leq c_2 \leq \frac{75}{2}$$

c_2 is the profit per bookcase. If this profit drops below 20, the carpenter should produce only tables. If the profit becomes greater than \$32.50 per bookcase, the carpenter should produce only bookcases. If the profit remains between these two values, then the carpenter should stick to the current optimal value of 12 tables and 15 bookcases.

Page 295, Question 3

Use the Curve Fitting Criterion to minimize the sum of the absolute deviations for the following models and data set:

- a. $y = ax$
- b. $y = ax^2$
- c. $y = ax^3$

First lets define our data:

```
x <- c(7,14,21,28,35,42)
y <- c(8,41,133,250,280,297)

data <- data.frame(x=x, y=y, x2=x^2, x3=x^3)
```

for each part of this question, I can define a cost function. The cost function is simply:

$$f(c) = \sum |y - ax^k|$$

Where k is the power we're raising it to, 1 in a, 2 in b, and 3 in c.

There are a few problems with this method. We need to define the interval beforehand, which means we need to have an idea of where the desired a lies. Lets just solve using the lm function to find some approximations of where we should be looking (we won't expect them to match, since with lm we're minimizing the sum of the squared errors, but it should give us a ballpark goal.

```
summary(lm(y~x+0,data=data))
```

```
##
## Call:
## lm(formula = y ~ x + 0, data = data)
##
## Residuals:
##      1      2      3      4      5      6
## -43.33 -61.66 -20.99  44.68  23.35 -10.98
##
## Coefficients:
##      Estimate Std. Error t value Pr(>|t|)
## x      7.3328      0.6276   11.68 8.07e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 41.91 on 5 degrees of freedom
## Multiple R-squared:  0.9647, Adjusted R-squared:  0.9576
## F-statistic: 136.5 on 1 and 5 DF, p-value: 8.071e-05
```

```
summary(lm(y~x2+0,data=data))
```



```
##
## Call:
## lm(formula = y ~ x2 + 0, data = data)
##
## Residuals:
##      1      2      3      4      5      6
## -2.1367  0.4532 41.7697 87.8127 26.5824 -67.9213
##
## Coefficients:
##      Estimate Std. Error t value Pr(>|t|)
## x2  0.20687    0.02326    8.893 0.000299 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 54.37 on 5 degrees of freedom
## Multiple R-squared:  0.9405, Adjusted R-squared:  0.9286
## F-statistic: 79.08 on 1 and 5 DF,  p-value: 0.0002993
```

```
summary(lm(y~x3+0,data=data))
```

```
##
## Call:
## lm(formula = y ~ x3 + 0, data = data)
##
## Residuals:
##      1      2      3      4      5      6
##  6.227 26.818 85.135 136.542 58.403 -85.919
##
## Coefficients:
##      Estimate Std. Error t value Pr(>|t|)
## x3 0.0051684  0.0009735   5.309  0.00317 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 86.54 on 5 degrees of freedom
## Multiple R-squared:  0.8494, Adjusted R-squared:  0.8192
## F-statistic: 28.19 on 1 and 5 DF,  p-value: 0.003168
```

All of these are between $[0,10]$, so I'll use that as my interval. I'm also setting a tolerance level as 0.01.

At the end of the while loop, the data frame solvetable will give me the values a and b, which give a range of possible values for a that is within my set tolerance.

a. $y = ax$

```
psi <- ((1+sqrt(5))/2)-1

cost <- function(a, data){
  sum(abs(data$y - a*(data$x)))
}

a <- 0
```

```

b <- 10
c1 <- a+(1-psi)*(b-a)
c2 <- a+psi*(b-a)
fc1 <- cost(c1,data)
fc2 <- cost(c2,data)

solvetable <- data.frame(a=a,b=b,c1=c1,c2=c2,fc1=fc1,fc2=fc2)

tolerance <- 0.01

t <- 1

while(t > tolerance){
  if(fc1<=fc2){
    b <- c2
    c1 <- a+(1-psi)*(b-a)
    c2 <- a+psi*(b-a)
    fc1 <- cost(c1,data)
    fc2 <- cost(c2,data)
    solvetable <- rbind(solvetable,c(a,b,c1,c2,fc1,fc2))
    t <- b-a
  } else{
    a <- c1
    c1 <- a+(1-psi)*(b-a)
    c2 <- a+psi*(b-a)
    fc1 <- cost(c1,data)
    fc2 <- cost(c2,data)
    solvetable <- rbind(solvetable,c(a,b,c1,c2,fc1,fc2))
    t <- b-a
  }
}

solvetable

```

```

##           a           b           c1           c2           fc1           fc2
## 1  0.000000 10.000000 3.819660 6.180340 509.9357 262.0643
## 2  3.819660 10.000000 6.180340 7.639320 262.0643 211.4257
## 3  6.180340 10.000000 7.639320 8.541020 211.4257 268.2328
## 4  6.180340 8.541020 7.082039 7.639320 199.7228 211.4257
## 5  6.180340 7.639320 6.737621 7.082039 220.5299 199.7228
## 6  6.737621 7.639320 7.082039 7.294902 199.7228 204.1929
## 7  6.737621 7.294902 6.950483 7.082039 207.1196 199.7228
## 8  6.950483 7.294902 7.082039 7.163346 199.7228 201.4303
## 9  6.950483 7.163346 7.031789 7.082039 201.9973 199.7228
## 10 7.031789 7.163346 7.082039 7.113096 199.7228 200.3750
## 11 7.031789 7.113096 7.062846 7.082039 200.0407 199.7228
## 12 7.062846 7.113096 7.082039 7.093902 199.7228 199.9719
## 13 7.062846 7.093902 7.074708 7.082039 199.5689 199.7228
## 14 7.062846 7.082039 7.070177 7.074708 199.5789 199.5689
## 15 7.070177 7.082039 7.074708 7.077508 199.5689 199.6277
## 16 7.070177 7.077508 7.072977 7.074708 199.5325 199.5689

```

Now lets just redefine our cost function and do the same thing for b and c:

b. $y = ax^2$

```
psi <- ((1+sqrt(5))/2)-1

cost <- function(a, data){
  sum(abs(data$y - a*(data$x^2)))
}

a <- 0
b <- 10
c1 <- a+(1-psi)*(b-a)
c2 <- a+psi*(b-a)
fc1 <- cost(c1,data)
fc2 <- cost(c2,data)

solvetable <- data.frame(a=a,b=b,c1=c1,c2=c2,fc1=fc1,fc2=fc2)

tolerance <- 0.01

t <- 1

while(t > tolerance){
  if(fc1<=fc2){
    b <- c2
    c1 <- a+(1-psi)*(b-a)
    c2 <- a+psi*(b-a)
    fc1 <- cost(c1,data)
    fc2 <- cost(c2,data)
    solvetable <- rbind(solvetable,c(a,b,c1,c2,fc1,fc2))
    t <- b-a
  } else{
    a <- c1
    c1 <- a+(1-psi)*(b-a)
    c2 <- a+psi*(b-a)
    fc1 <- cost(c1,data)
    fc2 <- cost(c2,data)
    solvetable <- rbind(solvetable,c(a,b,c1,c2,fc1,fc2))
    t <- b-a
  }
}

solvetable
```

##	a	b	c1	c2	fc1	fc2
## 1	0.0000000	10.0000000	3.8196601	6.1803399	16022.8644	26549.1356
## 2	0.0000000	6.1803399	2.3606798	3.8196601	9517.2711	16022.8644
## 3	0.0000000	3.8196601	1.4589803	2.3606798	5496.5933	9517.2711
## 4	0.0000000	2.3606798	0.9016994	1.4589803	3011.6778	5496.5933
## 5	0.0000000	1.4589803	0.5572809	0.9016994	1475.9155	3011.6778
## 6	0.0000000	0.9016994	0.3444185	0.5572809	526.7623	1475.9155
## 7	0.0000000	0.5572809	0.2128624	0.3444185	223.1277	526.7623
## 8	0.0000000	0.3444185	0.1315562	0.2128624	422.3910	223.1277
## 9	0.1315562	0.3444185	0.2128624	0.2631123	223.1277	285.5927

```
## 10 0.1315562 0.2631123 0.1818062 0.2128624 247.5555 223.1277
## 11 0.1818062 0.2631123 0.2128624 0.2320561 223.1277 223.2008
## 12 0.1818062 0.2320561 0.2009999 0.2128624 231.5670 223.1277
## 13 0.2009999 0.2320561 0.2128624 0.2201937 223.1277 219.8946
## 14 0.2128624 0.2320561 0.2201937 0.2247248 219.8946 217.8964
## 15 0.2201937 0.2320561 0.2247248 0.2275251 217.8964 216.6614
## 16 0.2247248 0.2320561 0.2275251 0.2292558 216.6614 217.5749
```

c. $y = ax^3$

```
psi <- ((1+sqrt(5))/2)-1

cost <- function(a, data){
  sum(abs(data$y - a*(data$x^3)))
}

a <- 0
b <- 10
c1 <- a+(1-psi)*(b-a)
c2 <- a+psi*(b-a)
fc1 <- cost(c1,data)
fc2 <- cost(c2,data)

solvetable <- data.frame(a=a,b=b,c1=c1,c2=c2,fc1=fc1,fc2=fc2)

tolerance <- 0.01

t <- 1

while(t > tolerance){
  if(fc1<=fc2){
    b <- c2
    c1 <- a+(1-psi)*(b-a)
    c2 <- a+psi*(b-a)
    fc1 <- cost(c1,data)
    fc2 <- cost(c2,data)
    solvetable <- rbind(solvetable,c(a,b,c1,c2,fc1,fc2))
    t <- b-a
  } else{
    a <- c1
    c1 <- a+(1-psi)*(b-a)
    c2 <- a+psi*(b-a)
    fc1 <- cost(c1,data)
    fc2 <- cost(c2,data)
    solvetable <- rbind(solvetable,c(a,b,c1,c2,fc1,fc2))
    t <- b-a
  }
}

solvetable
```

```
##      a      b      c1      c2      fc1      fc2
```

##	1	0	10.000000000	3.819660113	6.180339887	576764.2476	933847.7524
##	2	0	6.180339887	2.360679775	3.819660113	356074.5048	576764.2476
##	3	0	3.819660113	1.458980338	2.360679775	219680.7428	356074.5048
##	4	0	2.360679775	0.901699437	1.458980338	135384.7620	219680.7428
##	5	0	1.458980338	0.557280900	0.901699437	83286.9808	135384.7620
##	6	0	0.901699437	0.344418537	0.557280900	51088.7812	83286.9808
##	7	0	0.557280900	0.212862363	0.344418537	31189.1995	51088.7812
##	8	0	0.344418537	0.131556175	0.212862363	18890.5817	31189.1995
##	9	0	0.212862363	0.081306188	0.131556175	11289.6178	18890.5817
##	10	0	0.131556175	0.050249987	0.081306188	6591.9638	11289.6178
##	11	0	0.081306188	0.031056200	0.050249987	3688.6540	6591.9638
##	12	0	0.050249987	0.019193787	0.031056200	1897.1429	3688.6540
##	13	0	0.031056200	0.011862413	0.019193787	856.3900	1897.1429
##	14	0	0.019193787	0.007331374	0.011862413	461.0334	856.3900
##	15	0	0.011862413	0.004531039	0.007331374	401.0127	461.0334
##	16	0	0.007331374	0.002800336	0.004531039	585.4128	401.0127