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Tara Ram Mohan
Homework 6
OPER 527
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Question 1
(a) (10 points) Set up the problem using a penalty function. Clearly define your penalty functions.
# Given function: x^2 - 24*x + y^2 - 12*y + 200
f1 \leq function(x, y)
result1 < x^2 - 24*x + y^2 - 12*y + 200
 return (result1)
# Function to give me the maximum or 0
\max 0 \le \text{-function}(x)
result1 <- max( x, 0 )
 return(result1)
# Function to give me the minimum or 0
min0 \le function(x)
 result1 <- min(x, 0)
 return(result1)
}
# Penalty function
# Constraint 1: 2*x + 4*y \le 12
# Constraint 2: x + 3*y \le 15
# Constraint 3: x \ge 0
# Constraint 4: y \ge 0
p1 \le function(x, y)
c1 \le max0(-12 + 2*x + 4*y)
 c2 < -max0(-15 + x + 3*y)
 c3 \le min(x)
 c4 < -min(y)
 result1 <- mu1*(c1^2 + c2^2 + c3^2 + c4^2)
 return (result1)
# Function with penalty
p theta1 <- function(x, y){
 result1 <- f1(x, y) - p1(x, y)
 return (result1)
```

(b) (30 points) Use a gradient method to solve the penalty defined problem for  $\mu = 1$ , 10, and 100. Do the solutions differ across  $\mu$ ?

```
mu1 <- 100
mu2 <- 10
mu3 <- 1
# D/dx
p thetaprime 1x <-function(x, y){
result1 <-2*x - 24
c1 < -max0(-12 + 2*x + 4*y)
 c2 < -max0(-15 + x + 3*y)
 c3 \le min0(x)
 c4 \le min0(y)
 result1 <- result1 - mu1*(2*2*c1^4 + 1*2*c2^4 + 2*c3^5)
return(result1)
}
# D/dv
p_{thetaprime1y} < -function(x, y)
result1 <- 2*y - 12
c1 < -max0(-12 + 2*x + 4*y)
 c2 < -max0(-15 + x + 3*y)
 c3 \le min0(x)
 c4 \le min0(y)
 result1 <- result1 - mu1*(4*2*c1^4 + 3*2*c2^4 + 2*c4^5)
 return (result1)
# Gradient function
p g1 < -function(x, y)
 result1 <- c(p thetaprime1x(x, y), p thetaprime1y(x, y))
return(result1)}
x0 \le c(0, 0) #starting value for steepest ascent
step1 < -0.0001
count1 <- 1
error1 <-1
# Steepest ascent while loop
while(error 1 > 0.001)
x0hold <- x0
 x0 < -x0 + p \ g1(x0[1], x0[2])*step1
 p ghold <- p g1(x0[1], x0[2])
```

```
error1 <- t(p ghold)%*%(p ghold)
 count1 < -count1 + 1
#Penalty Solutions
# for mu3 = 1
p sol1 <- c(-1.69, -1.5)
\# for mu2 = 10
p sol10 <- c(-1.055, -0.929)
\# for mu1 = 100
p sol100 <- c(-0.661, -0.58)
Yes, the solutions differ across \mu.
(c) (10 points) Set up the problem using a barrier function. Clearly define your barrier functions.
mu1 <- 100
mu2 <- 10
mu3 <- 1
r = 1 \# barrier parameter
# Barrier function
# Constraint 1: 2*x + 4*y \le 12
# Constraint 2: x + 3*y \le 15
# Constraint 3: x \ge 0
# Constraint 4: y \ge 0
b1 \leq function(x, y)
 c1 < r*ifelse(is.nan(log(-2*x - 4*y + 12)), 0, log(-2*x - 4*y + 12))
 c2 < -r*ifelse(is.nan(log(-x - 3*y + 15)), 0, log(-x - 3*y + 15))
 c3 \le r*ifelse(is.nan(log(x)),0,log(x))
 c4 \le r*ifelse(is.nan(log(y)),0,log(y))
 result1 <- mu1*(c1 + c2 + c3 + c4)
 return (result1)
# Function with penalty
b theta1 <- function(x, y){
result1 <- f1(x, y) + b1(x, y)
 return(result1)
```

(d) (30 points) Use a gradient method to solve the barrier defined problem for  $\mu$  = 1, 10, and 100. Do the solutions differ across  $\mu$ .

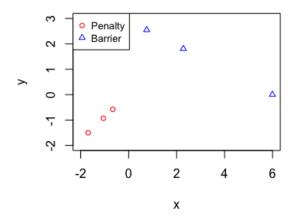
```
# D/dx
b thetaprime 1x <-function(x, y){
 result1 <- 2*x - 24
 c1 < r*ifelse(is.nan(log(-2*x - 4*y + 12)), 0, log(-2*x - 4*y + 12))
 c2 < -r*ifelse(is.nan(log(-x - 3*y + 15)), 0, log(-x - 3*y + 15))
 c3 \le r*ifelse(is.nan(log(x)),0,log(x))
 c4 \le r*ifelse(is.nan(log(y)), 0, log(y))
 result1 < - result1 + mu1*(-2*c1 - c2 + c3)
 return(result1)
}
# D/dy
b thetaprimely \leftarrow function(x, y){
result1 <- 2*y - 12
 c1 < r*ifelse(is.nan(log(-2*x - 4*y + 12)), 0, log(-2*x - 4*y + 12))
 c2 < -r*ifelse(is.nan(log(-x - 3*y + 15)), 0, log(-x - 3*y + 15))
 c3 \le r*ifelse(is.nan(log(x)), 0, log(x))
 c4 \le r*ifelse(is.nan(log(y)), 0, log(y))
 result1 <- result1 + mu1*(-4*c1 - 3*c2 + c4)
 return(result1)
# Gradient function
b g1 <- function(x, y){
 result1 <- c(b thetaprime1x(x, y), b thetaprime1y(x, y))
 return(result1)
}
x0 < c(0.5, 0.5)
step1 < -0.0001
count1 <- 1
error1 <- 1
# Steepest descent while loop
while(error1 > 0.0001) {
x0hold <- x0
 x0 \le x0 - step1*b g1(x0[1], x0[2])
 b ghold \leq- b g1(x0[1], x0[2])
 error1 <- t(b ghold)%*%(b ghold)
 count1 < -count1 + 1
#Barrier Solutions
```

```
# for mu3 = 1
b_sol1 <- c(5.99502, 0.00198)
# for mu2 = 10
b_sol10 <- c(2.28, 1.81)
# for mu1 = 100
b_sol100 <- c(0.752, 2.549)
```

Yes, the solutions differ across  $\mu$ .

(e) (10 points) Plot all of the solutions (x, y) on a single plot with colors that denote whether the Barrier Method or the Penalty method was used.

## **Penalty and Barrier Solutions**



## Question 2

(a) Write a simple Euler solver and program it.

lvsolve <- function(alpha, beta, S0, I0, R0, n) {

```
Sout 1 < c(S0)
 Iout1 <- c(I0)
 Rout1 <- c(R0)
 for (i in 2:n) {
  S0h \leftarrow Sout1[i-1]
  I0h <- Iout1[i-1]
  R0h \leq Rout1[i-1]
  Sout1[i] <- S0h - alpha*S0h*I0h
  Iout1[i] \le -alpha*S0h*I0h - beta*I0h + I0h
  Rout1[i] \le beta*I0h + R0h
 out1 <- data.frame(Sout1,Iout1,Rout1)
 names(out1) <- c("S","I","R")
 return(out1)
(b) Suppose we see the following data.
data1 <- matrix(c(
 0, 990, 10, 0,
 1, 986, 14, 0,
 2, 980, 19, 1,
 3, 973, 26, 1,
 4, 963, 36, 1,
 5, 949, 49, 2,
 6, 930, 67, 3,
 7, 905, 90, 5,
 8, 872, 120, 8,
 9, 830, 159, 11,
 10, 777, 207, 16,
 11, 713, 265, 22,
 12, 637, 333, 30,
 13, 552, 407, 41,
 14, 462, 483, 55,
 15, 373, 554, 73,
```

```
16, 290, 614, 96
), nrow = 17, byrow = T)
data1 <- data.frame(data1)
names(data1) <- c('Day', "S", "I", "R")
(c) Set up a least squares function with barrier functions for the constraints and program it.
SSLV <- function(alpha,beta,S0,I0,R0,n){
 # Estimated output
 euler out <- lvsolve(alpha,beta,S0,I0,R0,n)
 # Least squares calculation
 SS1 S <- sum((euler out S - data S)^2)
 SS1 I <- sum(( euler out$I - data1$I )^2)
 SS1 R <- sum(( euler outR - data1R)^2)
 SS1 \leftarrow SS1 S + SS1 I + SS1 R
# Add barrier functions for constaints
 SS1 <- SS1 + max( 0.0000001, 1/(alpha-0.00000001)^0.1 ) #changed
 SS1 <- SS1 + max( 0.0000001, 1/(beta-0.00000001)^0.1 ) #changed
return (SS1) #changed
(d) Use Gauss-Newton method to solve for a and b. You will likely need to use numerical derivatives.
# Numerical derivatives
dSSLV <- function(param1,S0,I0,R0,n){
h1 <- 0.0001
 alpha <- param1[1]
 beta <- param1[2]
# alpha parameter
 g1 \le SSLV(alpha+h1,beta,S0,I0,R0,n) - SSLV(alpha-h1,beta,S0,I0,R0,n)
 g1 < -(g1)/(2*n*h1)
 # beta parameter
 g2 \leq -SSLV(alpha,beta+h1,S0,I0,R0,n) - SSLV(alpha,beta-h1,S0,I0,R0,n)
 g2 < -(g2)/(2*n*h1)
 return(c(g1,g2))
```

```
# Hessian Matrix
hessian1 <-function(param1,S0,I0,R0,n){
h1 <- 0.00000001
 alpha <- param1[1]
 beta <- param1[2]
 matrix 11 \le SSLV(alpha+h1,beta,S0,I0,R0,n) + SSLV(alpha-h1,beta,S0,I0,R0,n)
 matrix11 <- matrix11 - 2*SSLV(alpha,beta,S0,I0,R0,n)
 matrix11 \le matrix11/(n*h1^2)
 matrix 22 \le SLV(alpha,beta+h1,S0,I0,R0,n) + SSLV(alpha,beta-h1,S0,I0,R0,n)
 matrix22 <- matrix22-2*SSLV(alpha,beta,S0,I0,R0,n)
 matrix22 \le matrix22/(n*h1^2)
 matrix12 <- SSLV(alpha+h1,beta+h1,S0,I0,R0,n) + SSLV(alpha-h1,beta-h1,S0,I0,R0,n)
 matrix12 <- matrix12 - SSLV(alpha-h1,beta+h1,S0,I0,R0,n)- SSLV(alpha+h1,beta-h1,S0,I0,R0,n)
 matrix 12 \le matrix 12/(4*n*h1^2)
 result1 <- matrix(c(matrix11,matrix12,
             matrix 12, matrix 22), nrow = 2, byrow = T)
 return(result1)
}
# Given starting values
S0 <- 990
I0 < -10
R0 < -0
n \le nrow(data1)
param1 <- c(0.00045,0.05) # Starting value
# Gausse-Newton Method
param1 <- param1 - solve( hessian1(param1,S0,I0,R0,n) )%*%dSSLV(param1,S0,I0,R0,n)
error1 = 1;
counter = 0;
while (error1 > 0.00001) {
x1hold <- param1
param1 <- param1 - solve( hessian1(param1,S0,I0,R0,n))%*%dSLV(param1,S0,I0,R0,n)
error1 <- sum( abs(x1hold - param1) )
 counter <- counter +1
```

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# Question 2 TEST -----
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```
# Plot given datapoints
plot(data1$Day,data1$S,
   xlab = "t", ylab = "",
   main = "Model of S, R, and I",
   xlim = c(0,18), ylim = c(0,1100),
   col = "magenta")
points(data1$Day,data1$I, col = "blue")
points(data1$Day,data1$R, col = "green")
# Plot estimated output
euler_out <- lvsolve(param1[1],param1[2],S0,I0,R0,n)</pre>
lines(euler out$S, col = "magenta")
lines(euler out$I, col = "blue")
lines(euler_out$R, col = "green")
# Legend
legend("left", legend=c("S(t)", "I(t)", "R(t)"),
    col=c("magenta", "blue", "green"), lty=1:2, cex=0.8)
```

## Model of S, R, and I

