

Third Assignment

(Make function for Multiple Linear Regression Analysis

MyReg( ) )

# Exploratory Data Analysis & Statistical Consulting

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## Function MyReg()

주어진 자료에서 반응변수와 설명변수들에 관한 다중회귀분석(Multiple Linear Regression)을 실시하여 해당하는 통계량들을 계산하여 화면에 보여주고, 회귀분석 결과에 따른 산점도, 실제 반응변수의 값과 회귀모형에 의해 예측된 값을 비교하는 그림, 예측치와 표준화된 잔차에 대한 그래프, 그리고 표준화된 잔차의 절대값에 대하여 퍼진 정도를 알아보는 그래프를 그려서 보여준다.

### # Case 1

input

```
N <- 100
set.seed(1234)
tX1 <- rnorm(N,0,1)
tX2 <- rnorm(N,3,1)
tX3 <- rgamma(N,1,3)
tX4 <- sample(c(0,1),N,replace=T)
tX5 <- sample(1:3,N,replace=T,prob=c(1,2,3)/6)
tX6 <- rbinom(N,4,0.3)
tX <- cbind(rep(1,length(tX1)),tX1,tX2,tX3,tX4,tX5,tX6)
tbeta <- (0:6)/5
ty <- as.vector(tX*%tbeta)+rnorm(N,0,2)
dat <- data.frame(y=ty, age=tX1, height=tX2, weight=tX3, smoking=tX4,
                  therapy=tX5, surgery=tX6)

MyReg <- function(data) {
  n <- nrow(data) ; p <- ncol(data)-1
  y <- as.matrix(data)[,1] ; X <- cbind(rep(1,n),as.matrix(data)[,-1])
  beta <- (solve(t(X)%*%X))%*%(t(X))%*%(y)
  pred <- X%*%beta
  residuals <- y-pred ; sig2 <- (t(residuals)%*%residuals)/(n-p-1)
  H <- (X)%*%(solve(t(X)%*%X))%*%(t(X))
  stdresid <- rep(0,n)
  for(i in 1:n) {
    stdresid[i] <- residuals[i]/sqrt(sig2*(1-H[i,i]))
  }
  SST <- sum((y-mean(y))^2) ; SSE <- sum((y-pred)^2) ; SSR <- SST - SSE
  MSR <- SSR/(p) ; MSE <- SSE/(n-p-1) ; F <- MSR/MSE
  P.value <- 1-pf(F,p,n-p-1) ; Rsq <- SSR/SST
  C <- solve(t(X)%*%X)
  std.error <- rep(0,p+1)
  for(i in 1:(p+1)) {
    std.error[i] <- sqrt(sig2*C[i,i])
  }
  t.value <- rep(0,p+1)
  for(i in 1:(p+1)) {
```

```

    t.value[i] <- beta[i]/std.error[i]
  }
  Pr.t <- rep(0,(p+1))
  for(i in 1:(p+1)) {
    Pr.t[i] <- 2*(1-pt(abs(t.value[i]), n-p-1))
  }
  args <- list(Data=data, Beta=beta, Sigma2=sig2, Predict=pred,
               Residuals=residuals, Standard.Residuals=stdresid, SSR=SSR,
               SSE=SSE, F.Statistic=F, P.value=P.value)

  aster <- function(p) {
    if (p >= 0.05 & p < 0.1) "."
    else if (p >= 0.01 & p < 0.05) "*"
    else if (p >= 0.001 & p < 0.1) "***"
    else if (p < 0.001) "****"
    else " "
  }

  cat("\n == ANALYSIS OF VARIANCE == \n\n",
      encodeString(c("Source", " df", " SS", " MS", " F", " P-value"),
                   width=8, justify="right"), "\n",
      "-----", "\n",
      encodeString(c("Regression", p, round(SSR,2), round(MSR,2), round(F,2),
                    round(P.value,2)), width=8, justify="right"),
      aster(P.value), "\n", encodeString(c(" Error", n-p-1, round(SSE,2),
                    round(MSE,2)), width=8, justify="right"),
      "\n", encodeString(c(" Total", n-1, round(SST,2)), width=8,
                    justify="right"),
      "\n",
      "-----", "\n",
      "Estimated error variance :", round(sig2,4), "\n", "R-squares :", round(Rsq,4),
      "\n\n")

  ind.name <- colnames(data)
  ind.name[1] <- "(Intercept)"
  test.mat <- cbind(ind.name, round(beta,4), round(std.error,4), round(t.value,4),
                    round(Pr.t,4))
  cat("\n\n\n== PARAMETER ESTIMATES ==\n\n",
      encodeString(c(" ", "Estimate", "Std.Error", "t value", "Pr(>|t|)"), width=11,
                   justify="right"),
      "\n")
  for(i in 1:nrow(test.mat)) {
    cat(encodeString(test.mat[i,], width=11, justify="right"), aster(Pr.t[i]), "\n")
  }

  cat("\n", "===Various Statistics in Multiple Regression===", "\n\n")
  print(args)

  outlier1 <- rep(0,n)
  outlier2 <- rep(0,n)
  std.sign <- rep(0,n)

```

```

for(i in 1:n) {
  outlier1[i] <- ifelse(abs(stdresid[i]) > 2, "out", "in")
  outlier2[i] <- ifelse(abs(stdresid[i]) > 2, 2, 1)
  std.sign[i] <- ifelse(stdresid[i] > 0, 24, 20)
}
abs.stdresid <- abs(stdresid)
ex1 <- as.data.frame(cbind(number=(1:n), stdresid=stdresid,
                           pred=pred, y=y,
                           outlier1=outlier1, outlier2=outlier2,
                           std.sign=std.sign, abs.stdresid=abs.stdresid))
ex1_1 <- as.matrix(ex1[,1:4,8])
ex1_1 <- as.numeric(ex1_1)
ex1_1 <- matrix(ex1_1,nrow=n,ncol=5)
colnames(ex1_1) <- c("number","stdresid","pred","y","abs.stdresid")
ex2 <- as.data.frame(ex1_1)
ex2 <- ex2[abs.stdresid > 2,]

pred1 <- as.vector(pred) ; abs.stdresid1 <- as.vector(abs.stdresid)
ord <- order(pred1,abs.stdresid1)
re1 <- as.data.frame(t(rbind(pred1,abs.stdresid1)[,ord]))

KDE <- function(x) {
  w <- rep(0,n)
  for (i in 1:n) {
    w[i] <- exp(-(pred[i]-x)^2)/sqrt(2*pi)
  }
  ans <- sum(w*abs(stdresid))/sum(w)
  return(ans)
}
d1 <- seq(min(re1$pred1),max(re1$pred1),0.01)
d2 <- rep(0,length(d1))
for (j in 1:length(d1)) {
  d2[j] <- KDE(d1[j])
}
d3 <- re1$pred1
d4 <- rep(0,length(d3))
for (j in 1:length(d3)) {
  d4[j] <- mean(re1$abs.stdresid1[j:(j+9)],na.rm=TRUE)
}

pairs(data, main="Scatterplot matrix from given data", cex=outlier2,
      pch=21, bg=c("yellow","red")[unclass(ex1$outlier1)])
dev.new()
plot(y~pred, main="Dependent variable VS Fitted values",
     xlab="Fitted response values",
     ylab="Response variables", cex=abs(stdresid)*7, pch=20,
     col=factor(outlier2, labels=c("black", "red")))
abline(0,1,col="blue",lwd=2)
for (i in 1:nrow(ex2)) {
  text(ex2$pred[i], ex2$y[i], ex2$number[i], col="light blue")
}

```

```

dev.new()
par(fig=c(0,0.7,0,1), new=FALSE)
plot(stdresid~pred, main="Standard residuals VS predicted values",
      xlab="Fitted response values", ylab="Studentized residuals",
      cex=abs(stdresid)*3, pch=20,
      col=factor(outlier2, labels=c("black", "orange")))
abline(h=c(-2,0,2),col="blue",lty=c(2,1,2))
for (i in 1:nrow(ex2)) {
  text(ex2$pred[i], ex2$stdresid[i], ex2$number[i], col="yellow")
}
par(fig=c(0.7,1,0,1),new=TRUE)
boxplot(stdresid,col="green")
for (i in 1:6) {
  text(0.7,summary(stdresid)[i],round(summary(stdresid)[i],3),col="purple")
}
dev.new()
par(fig=c(0,1,0,1),new=FALSE)
plot(abs.stdresid~pred,
      main="Absolute Standard residuals VS predicted values",
      xlab="Fitted response values",
      ylab="Absolute studentized residuals", cex=1.5, pch=std.sign)
legend("topright",c("positive","negative"),pch=c(24,20))
lines(d1,d2,lty=3,lwd=2,col="red")
lines(d3,d4,lty=2,lwd=2,col="blue")
legend("topleft",c("KDE","MA10"),lty=c(3,2),lwd=2,col=c("red","blue"))
dev.new()
qqnorm(stdresid) : qqline(stdresid)
test.res <- shapiro.test(stdresid)
print(test.res)

return(args)
}

res1 <- MyReg(dat)

str(res1)

attributes(res1)

```

output

```
== ANALYSIS OF VARIANCE ==
```

Source	df	SS	MS	F	P-value
Regression	6	195.3	32.55	8.72	0 ***
Error	93	347.11	3.73		
Total	99	542.41			

-----

Estimated error variance : 3.7323

R-squares : 0.3601

== PARAMETER ESTIMATES ==

	Estimate	Std.Error	t value	Pr(> t )
(Intercept)	-0.7866	1.182	-0.6655	0.5074
age	0.0198	0.1951	0.1012	0.9196
height	0.5986	0.1953	3.0653	0.0028 **
weight	-0.0803	0.6073	-0.1322	0.8951
smoking	0.2927	0.4105	0.713	0.4776
therapy	1.2801	0.2794	4.5809	0 ***
surgery	1.0944	0.2018	5.4219	0 ***

==Various Statistics in Multiple Regression==

\$Data

	y	age	height	weight	smoking	therapy	surgery
1	7.6185092	-1.207065749	3.4145235	0.300656147	1	2	3
2	11.1395846	0.277429242	2.5252815	0.371353870	0	3	2
3	5.7838752	1.084441177	3.0659935	0.213260667	0	2	1
4	6.2143115	-2.345697703	2.4975222	0.372750095	1	2	1
5	7.7605163	0.429124689	2.1740014	0.248226012	0	2	1
6	7.7740276	0.506055892	3.1669893	0.394101281	0	3	1
7	2.0073677	-0.574739960	2.1037354	0.883828859	0	2	1
8	6.8965796	-0.546631856	3.1681854	0.531965694	1	2	1
9	7.7153296	-0.564451999	3.3549683	0.174454546	0	3	3
10	8.7708591	-0.890037829	2.9478949	0.007488234	1	1	2
11	5.1149679	-0.477192700	2.8040654	0.626116131	0	3	1
12	1.3942546	-0.998386445	2.3509302	1.179623070	1	1	1
13	8.0460598	-0.776253895	1.8902328	0.102581133	1	2	3
14	6.5964155	0.064458817	3.8492742	0.224677931	0	3	1
15	7.1039366	0.959494059	3.0223625	0.563909112	0	3	2
16	1.4979346	-0.110285494	3.8311406	0.167726840	0	2	0
17	3.2428745	-0.511009506	1.7557121	0.201034863	0	3	1
18	6.2127608	-0.911195417	3.1690264	0.254401211	1	3	1
19	8.1941169	-0.837171680	3.6731663	0.693440641	0	2	2
20	6.0718540	2.415835178	2.9737236	0.213521547	1	3	1
21	3.4667472	0.134088220	2.8086078	0.098975684	1	2	1
22	0.4867931	-0.490685897	2.2180934	0.176831684	1	3	1
23	4.7385541	-0.440547872	5.0581620	0.024831943	1	1	0
24	4.5970284	0.459589441	3.7505015	0.250382061	0	2	0
25	7.8639016	-0.693720247	4.8242083	0.757634302	0	3	0
26	6.9645135	-1.448204910	3.0800596	0.147808868	0	3	0
27	5.0848131	0.574755721	2.3685907	0.180896373	0	3	1
28	8.7539894	-1.023655723	1.4867119	0.278347274	0	2	3
29	8.3501663	-0.015138300	2.3639002	0.008344596	1	3	3
30	5.3399869	-0.935948601	3.2263015	0.315793550	0	3	2
31	5.1378487	1.102297546	4.0136903	0.013843684	1	3	1
32	5.0961565	-0.475593079	3.2527501	0.640095477	0	3	1

33	5.0096718	-0.709440038	1.8280517	0.275633695	1	3	0
34	4.4533125	-0.501258061	3.6687143	0.234801962	0	3	0
35	4.9573960	-1.629093469	1.3498991	0.235832515	0	3	1
36	5.3893573	-1.167619262	2.6341478	0.285561219	1	2	1
37	7.7728607	-2.180039649	2.6838817	0.510195492	0	3	2
38	3.9803126	-1.340993192	1.0517540	0.290288645	0	3	1
39	7.9956412	-0.294293859	3.9200575	0.359636780	0	3	0
40	5.5369194	-0.465897540	2.3771284	0.006432923	1	3	0
41	3.2315194	1.449496265	2.6659634	0.770510833	1	3	0
42	6.7268347	-1.068642724	4.3951479	0.226339440	0	2	3
43	3.5572367	-0.855364634	3.6366744	0.403122869	0	1	0
44	3.9056203	-0.280623002	2.8915683	1.045380347	1	1	1
45	5.7657251	-0.994340076	3.5137628	0.115163135	1	3	0
46	6.8175870	-0.968514318	3.3992718	1.304649095	1	2	1
47	6.4152961	-1.107318193	4.6628564	0.274436277	0	2	0
48	4.8616932	-1.251985886	3.2758934	0.175208053	1	3	1
49	5.4079627	-0.523828119	3.5062726	0.239968258	1	2	1
50	4.1239824	-0.496849957	3.3475520	0.032860857	1	2	1
51	5.3127850	-1.806031257	2.6227624	0.465253488	0	3	3
52	1.2245240	-0.582075925	3.0976195	0.999719975	0	2	1
53	6.4538941	-1.108889624	4.6387446	0.048526795	0	2	2
54	7.2596023	-1.014962009	2.1244075	0.016168668	0	3	2
55	7.5936876	-0.162309524	3.1217600	0.165215369	1	3	1
56	1.5583003	0.563055819	4.3621307	0.468037451	0	1	1
57	3.0788377	1.647817473	2.7653789	1.170797202	0	2	1
58	0.4581519	-0.773353424	1.9466172	0.527473505	1	1	0
59	3.3099758	1.605909629	2.1302164	0.523417265	0	2	1
60	5.2972146	-1.157808548	2.6098730	0.240719133	1	2	1
61	7.4377150	0.656588464	2.1526499	0.093950833	1	2	2
62	5.4576686	2.548991071	2.7393606	0.136985760	1	1	3
63	-1.9376925	-0.034760390	2.5855803	0.939535162	0	1	1
64	9.2465962	-0.669633580	2.8169492	0.243493845	1	3	2
65	7.3245442	-0.007604756	3.4070561	0.554455368	0	2	2
66	3.3409581	1.777084448	3.6246331	0.229048919	1	2	0
67	5.8706458	-1.138607737	4.6782057	0.290679589	1	3	2
68	6.0322781	1.367827179	2.9313063	0.310287531	1	3	1
69	4.7213034	1.329564791	2.6791601	1.945855520	0	2	0
70	7.5783331	0.336472797	4.4710057	0.127057828	1	1	2
71	7.6815165	0.006892838	4.7043294	0.030359993	0	3	0
72	4.4577125	-0.455468738	3.0432440	0.048792384	1	3	1
73	7.7402535	-0.366523933	2.6673427	1.275292553	0	3	2
74	9.7061666	0.648286568	1.1777646	0.395085718	1	3	2
75	3.7923051	2.070270861	4.4112624	0.694050412	0	3	0
76	2.3613461	-0.153398412	2.1624176	0.287642700	1	1	0
77	3.1786706	-1.390700947	1.8762372	0.082086647	1	3	2
78	5.1042255	-0.723581777	6.0437659	0.760985052	1	1	0
79	5.2019065	0.258261762	3.2350213	0.231048529	1	3	1
80	2.6156518	-0.317059115	2.9667414	0.237070641	0	2	1
81	5.2341128	-0.177789958	0.2677805	0.204189201	1	1	4
82	6.4998278	-0.169994077	2.9002094	0.474560230	1	1	2
83	-0.9153396	-1.372301886	3.9760317	0.489410169	1	1	0

84	5.9297626	-0.173787170	3.4138689	0.083323763	0	3	2
85	2.6109900	0.850232257	3.9123222	0.172416743	1	2	0
86	6.4501599	0.697608712	4.9837322	0.068972247	1	3	2
87	8.4683504	0.549997351	4.1691085	0.569063747	0	3	2
88	4.8777452	-0.402731975	2.4912630	0.430463151	0	3	0
89	7.8100242	-0.191593770	3.7041802	0.301186011	1	2	3
90	5.2398306	-1.194527880	2.8015837	0.138446892	0	2	1
91	5.2926870	-0.053158819	2.4619292	0.537928137	1	2	2
92	2.6914147	0.255196001	0.1442413	0.048790497	0	3	3
93	7.0006510	1.705964007	2.2103531	0.777457371	0	3	0
94	5.5954422	1.001513252	3.4878146	0.102332856	0	2	2
95	7.2981776	-0.495583443	5.1680325	0.767098073	0	2	3
96	4.8584337	0.355550297	3.5006946	0.120016478	1	1	1
97	7.0215120	-1.134608044	3.6202102	0.780808501	1	3	0
98	8.6139680	0.878203627	2.0340968	0.287524310	0	3	3
99	7.4413886	0.972916753	3.1626547	0.545354279	1	3	0
100	3.9999188	2.121117105	0.9217625	0.191202086	0	3	0
\$Beta							
	[.1]						
	-0.78660499						
age	0.01975009						
height	0.59862082						
weight	-0.08026628						
smoking	0.29271521						
therapy	1.28005300						
surgery	1.09439426						
\$Sigma2							
	[.1]						
[1,]	3.732336						
\$Predict							
	[.1]						
[1,]	7.345432						
[2,]	6.729701						
[3,]	4.707563						
[4,]	4.579432						
[5,]	4.157849						
[6,]	6.022136						
[7,]	4.044942						
[8,]	5.003657						
[9,]	8.319940						
[10,]	4.721444						
[11,]	5.766840						
[12,]	3.173471						
[13,]	6.457367						
[14,]	6.435443						
[15,]	7.025279						
[16,]	4.051261						
[17,]	5.172725						



[18.]	6.299293
[19.]	6.088929
[20.]	6.251371
[21.]	4.836605
[22.]	5.744576
[23.]	3.803390
[24.]	4.007609
[25.]	5.866912
[26.]	4.856876
[27.]	5.562668
[28.]	5.904101
[29.]	8.043563
[30.]	7.129841
[31.]	6.864001
[32.]	6.034341
[33.]	4.404443
[34.]	5.220976
[35.]	4.904922
[36.]	4.691485
[37.]	6.764963
[38.]	4.727765
[39.]	5.365503
[40.]	4.759550
[41.]	4.908952
[42.]	7.648438
[43.]	2.621186
[44.]	3.522059
[45.]	5.420799
[46.]	5.071638
[47.]	4.520886
[48.]	6.362891
[49.]	5.229931
[50.]	5.152074
[51.]	7.833764
[52.]	4.630455
[53.]	6.713343
[54.]	6.492714
[55.]	6.292947
[56.]	4.172657
[57.]	4.461878
[58.]	1.893837
[59.]	4.132791
[60.]	4.680746
[61.]	5.549052
[62.]	5.748532
[63.]	3.059525
[64.]	7.188573
[65.]	5.957170
[66.]	4.252710
[67.]	8.289710
[68.]	6.197514

[69.] 3.247374  
[70.] 5.647836  
[71.] 5.867363  
[72.] 6.249501  
[73.] 6.729468  
[74.] 6.221184  
[75.] 5.679407  
[76.] 2.054514  
[77.] 6.624157  
[78.] 4.328715  
[79.] 6.363770  
[80.] 4.618558  
[81.] 5.304138  
[82.] 4.669629  
[83.] 3.099912  
[84.] 7.275835  
[85.] 4.411167  
[86.] 8.526665  
[87.] 7.703244  
[88.] 4.502370  
[89.] 7.538839  
[90.] 4.510277  
[91.] 5.684539  
[92.] 6.424207  
[93.] 4.348007  
[94.] 6.061734  
[95.] 8.079016  
[96.] 3.973535  
[97.] 5.428321  
[98.] 7.548656  
[99.] 5.214942  
[100.] 3.631885

\$Residuals

[,1]

[1.] 0.27307748  
[2.] 4.40988390  
[3.] 1.07631221  
[4.] 1.63487922  
[5.] 3.60266744  
[6.] 1.75189198  
[7.] -2.03757453  
[8.] 1.89292234  
[9.] -0.60461022  
[10.] 4.04941549  
[11.] -0.65187166  
[12.] -1.77921650  
[13.] 1.58869294  
[14.] 0.16097255  
[15.] 0.07865770  
[16.] -2.55332600

[17.] -1.92985078  
[18.] -0.08653188  
[19.] 2.10518761  
[20.] -0.17951677  
[21.] -1.36985822  
[22.] -5.25778252  
[23.] 0.93516377  
[24.] 0.58941938  
[25.] 1.99698954  
[26.] 2.10763788  
[27.] -0.47785447  
[28.] 2.84988813  
[29.] 0.30660316  
[30.] -1.78985430  
[31.] -1.72615274  
[32.] -0.93818463  
[33.] 0.60522834  
[34.] -0.76766374  
[35.] 0.05247420  
[36.] 0.69787267  
[37.] 1.00789823  
[38.] -0.74745238  
[39.] 2.63013818  
[40.] 0.77736946  
[41.] -1.67743265  
[42.] -0.92160296  
[43.] 0.93605032  
[44.] 0.38356091  
[45.] 0.34492631  
[46.] 1.74594924  
[47.] 1.89440971  
[48.] -1.50119816  
[49.] 0.17803138  
[50.] -1.02809195  
[51.] -2.52097847  
[52.] -3.40593097  
[53.] -0.25944882  
[54.] 0.76688853  
[55.] 1.30074046  
[56.] -2.61435704  
[57.] -1.38304001  
[58.] -1.43568476  
[59.] -0.82281543  
[60.] 0.61646826  
[61.] 1.88866260  
[62.] -0.29086317  
[63.] -4.99721750  
[64.] 2.05802371  
[65.] 1.36737425  
[66.] -0.91175167  
[67.] -2.41906393

[68.] -0.16523549  
 [69.] 1.47392896  
 [70.] 1.93049729  
 [71.] 1.81415371  
 [72.] -1.79178830  
 [73.] 1.01078597  
 [74.] 3.48498284  
 [75.] -1.88710167  
 [76.] 0.30683238  
 [77.] -3.44548652  
 [78.] 0.77551044  
 [79.] -1.16186342  
 [80.] -2.00290589  
 [81.] -0.07002558  
 [82.] 1.83019893  
 [83.] -4.01525198  
 [84.] -1.34607257  
 [85.] -1.80017661  
 [86.] -2.07650540  
 [87.] 0.76510680  
 [88.] 0.37537499  
 [89.] 0.27118489  
 [90.] 0.72955358  
 [91.] -0.39185245  
 [92.] -3.73279184  
 [93.] 2.65264428  
 [94.] -0.46629188  
 [95.] -0.78083813  
 [96.] 0.88489860  
 [97.] 1.59319079  
 [98.] 1.06531237  
 [99.] 2.22644678  
 [100.] 0.36803345

\$Standard.Residuals

[1] 0.14621932 2.32114970 0.57153929 0.87974071 1.90940601 0.92011018 -1.08413684  
 [8] 0.99306711 -0.32351339 2.17266615 -0.34288737 -0.98076112 0.84919099 0.08485406  
 [15] 0.04173664 -1.36122109 -1.02336295 -0.04573283 1.11549389 -0.09804247 -0.71981894  
 [22] -2.77953994 0.51186661 0.31366989 1.07755297 1.13136116 -0.25226494 1.53626742  
 [29] 0.16453837 -0.94461232 -0.92611157 -0.49376898 0.32502459 -0.40735099 0.02834328  
 [36] 0.36761431 0.54242613 -0.40399631 1.39370414 0.41608024 -0.91057152 -0.49983195  
 [43] 0.50878722 0.20754560 0.18421732 0.96496966 1.01878534 -0.79633830 0.09331282  
 [50] -0.54082447 -1.36291393 -1.81089345 -0.13990659 0.40781113 0.68549359 -1.41232270  
 [57] -0.75325283 -0.78080575 -0.44049989 0.32484902 1.00074739 -0.16549547 -2.69924865  
 [64] 1.08900085 0.72007226 -0.49087686 -1.31213965 -0.08811307 0.87626859 1.04331695  
 [71] 0.97856093 -0.94605065 0.55642828 1.87485168 -1.03107986 0.16624905 -1.84199828  
 [78] 0.43186179 -0.61288542 -1.05455478 -0.04005917 0.97389414 -2.16611993 -0.71160574  
 [85] -0.95776832 -1.13700689 0.40890568 0.19899248 0.14514993 0.38777677 -0.20652954  
 [92] -2.06344447 1.43414533 -0.24914333 -0.43376220 0.47286605 0.86279698 0.57223897  
 [99] 1.19066657 0.20592839

\$SSR

[1] 195.2996

\$SSE

[1] 347.1072

\$F.Statistic

[1] 8.721064

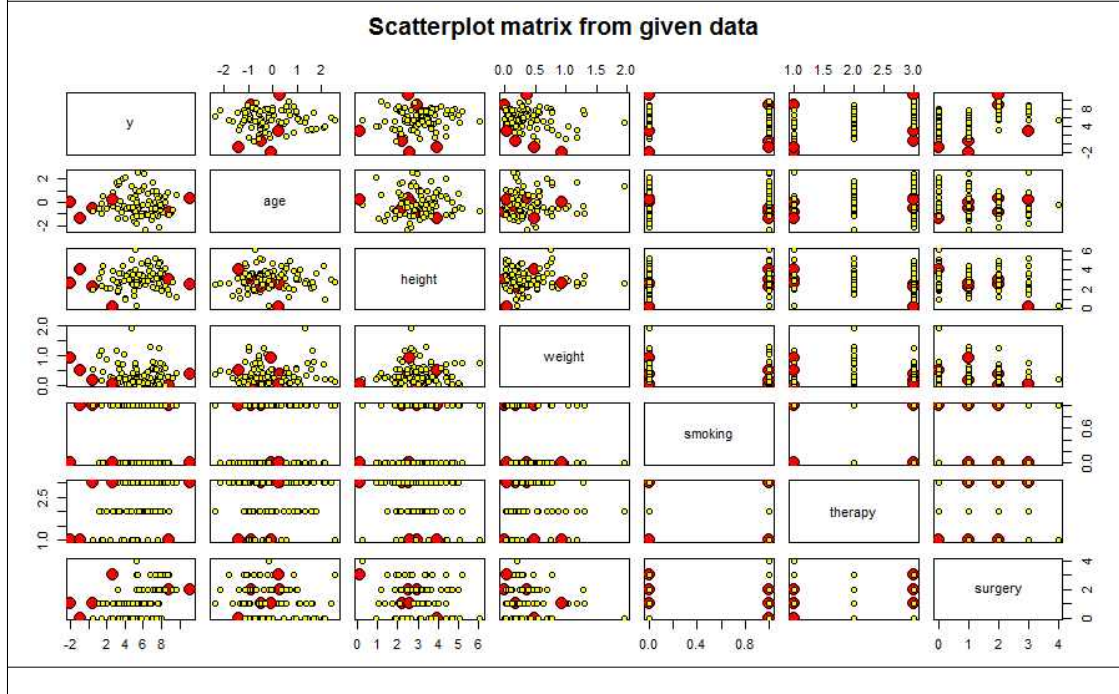
\$P.value

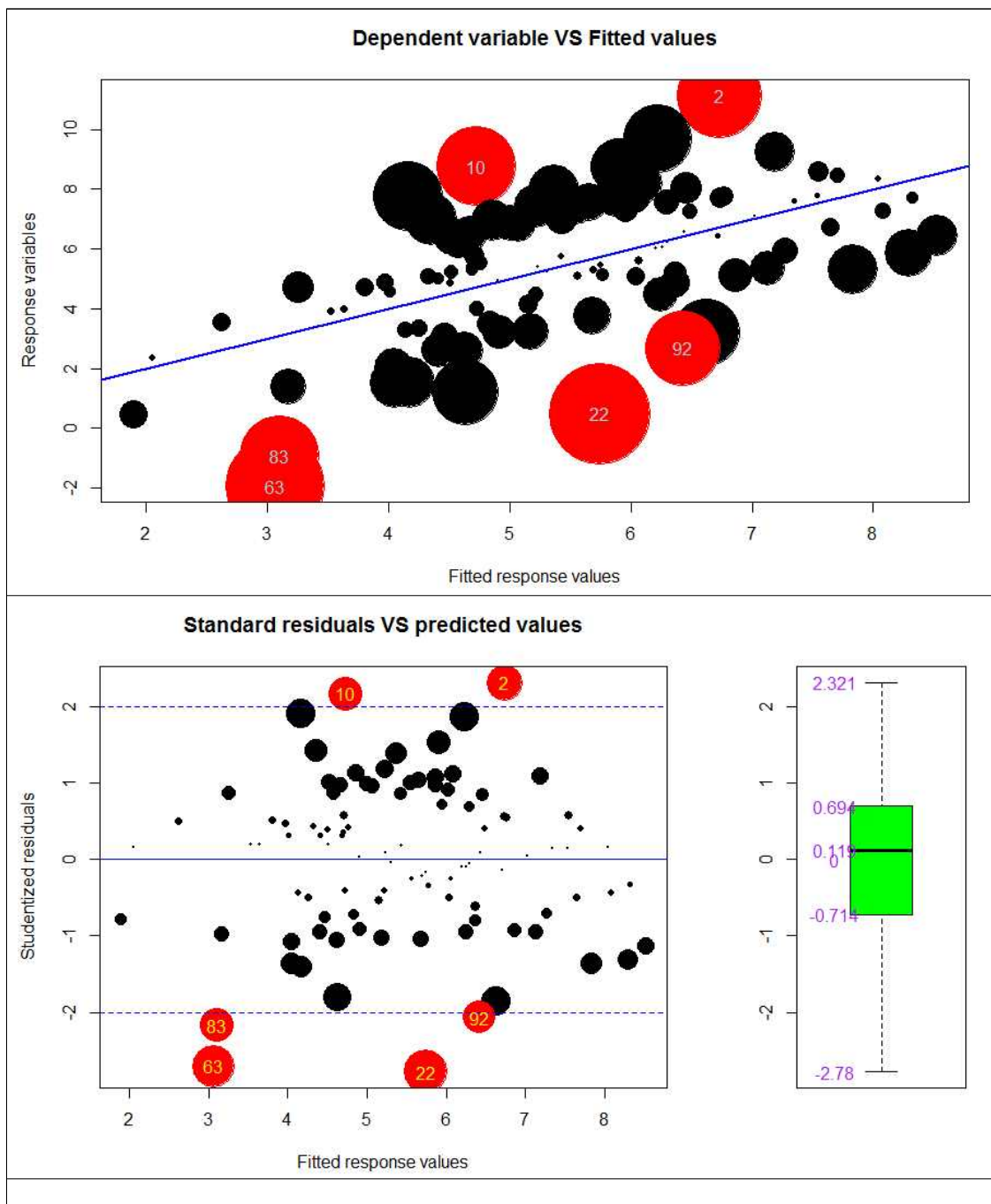
[1] 1.556127e-07

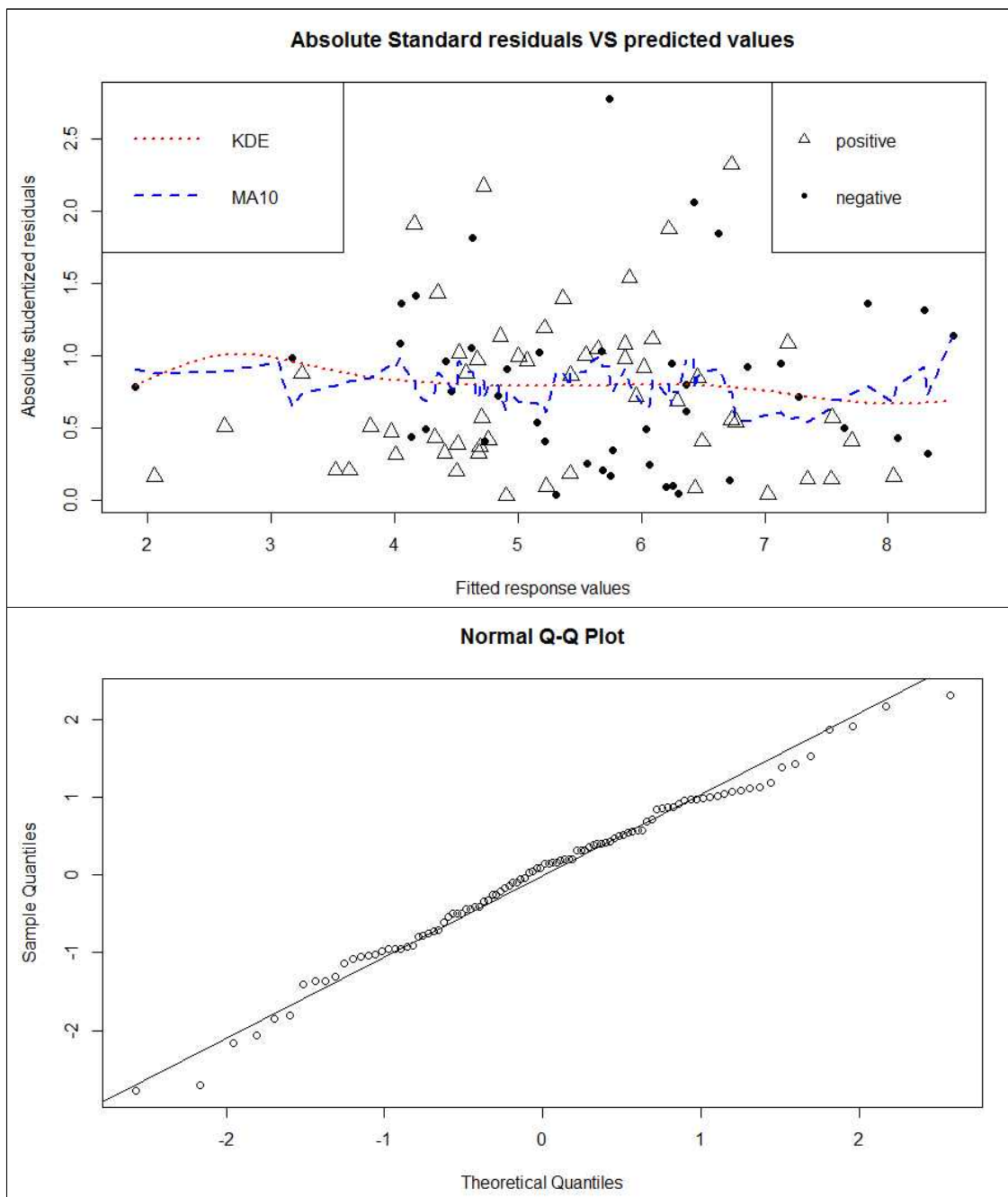
Shapiro-Wilk normality test

data: stdresid

W = 0.9875, p-value = 0.4694







## # Case 2

input

```
N <- 100
set.seed(1234)
tX1 <- rnorm(N,0,1)
tX2 <- rnorm(N,3,1)
tX3 <- rgamma(N,1,3)
tX4 <- sample(c(0,1),N,replace=T)
tX5 <- sample(1:3,N,replace=T,prob=c(1,2,3)/6)
tX6 <- rbinom(N,4,0.3)
tX <- cbind(rep(1,length(tX1)),tX1,tX2,tX3,tX4,tX5,tX6)
tbeta <- (0:6)/5
ty <- as.vector(tX*%tbeta)+rnorm(N,0,2)
dat <- data.frame(y=ty, age=tX1, height=tX2, weight=tX3, smoking=tX4,
                  therapy=tX5, surgery=tX6)

MyReg <- function(data) {
  n <- nrow(data) ; p <- ncol(data)-1
  y <- as.matrix(data)[,1] ; X <- cbind(rep(1,n),as.matrix(data)[,-1])
  beta <- (solve(t(X)%*%X))%*%(t(X))%*%(y)
  pred <- X%*%beta
  residuals <- y-pred ; sig2 <- (t(residuals)%*%residuals)/(n-p-1)
  H <- (X)%*%(solve(t(X)%*%X))%*%(t(X))
  stdresid <- rep(0,n)
  for(i in 1:n) {
    stdresid[i] <- residuals[i]/sqrt(sig2*(1-H[i,i]))
  }
  SST <- sum((y-mean(y))^2) ; SSE <- sum((y-pred)^2) ; SSR <- SST - SSE
  MSR <- SSR/(p) ; MSE <- SSE/(n-p-1) ; F <- MSR/MSE
  P.value <- 1-pf(F,p,n-p-1) ; Rsq <- SSR/SST
  C <- solve(t(X)%*%X)
  std.error <- rep(0,p+1)
  for(i in 1:(p+1)) {
    std.error[i] <- sqrt(sig2*C[i,i])
  }
  t.value <- rep(0,p+1)
  for(i in 1:(p+1)) {
    t.value[i] <- beta[i]/std.error[i]
  }
  Pr.t <- rep(0,(p+1))
  for(i in 1:(p+1)) {
    Pr.t[i] <- 2*(1-pt(abs(t.value[i]), n-p-1))
  }
  args <- list(Data=data, Beta=beta, Sigma2=sig2, Predict=pred,
              Residuals=residuals, Standard.Residuals=stdresid, SSR=SSR,
              SSE=SSE, F.Statistic=F, P.value=P.value)

  aster <- function(p) {
```



```

    if (p >= 0.05 & p < 0.1) "."
    else if (p >= 0.01 & p < 0.05) "*"
    else if (p >= 0.001 & p < 0.1) "***"
    else if (p < 0.001) "***"
    else " "
  }

cat("\n == ANALYSIS OF VARIANCE == \n\n",
    encodeString(c("Source", " df", " SS", " MS", " F", " P-value",
        width=8, justify="right"), "\n",
    "-----", "\n",
    encodeString(c("Regression", p, round(SSR,2), round(MSR,2), round(F,2),
        round(P.value,2)), width=8, justify="right"),
    aster(P.value), "\n", encodeString(c(" Error", n-p-1, round(SSE,2),
        round(MSE,2)), width=8, justify="right"),
    "\n", encodeString(c(" Total", n-1, round(SST,2)), width=8,
        justify="right"),
    "\n",
    "-----", "\n",
    "Estimated error variance :", round(sig2,4), "\n", "R-squares :", round(Rsq,4),
    "\n\n")

ind.name <- colnames(data)
ind.name[1] <- "(Intercept)"
test.mat <- cbind(ind.name, round(beta,4), round(std.error,4), round(t.value,4),
    round(Pr.t,4))
cat("\n\n\n== PARAMETER ESTIMATES ==\n\n",
    encodeString(c(" ", "Estimate", "Std.Error", "t value", "Pr(>|t|)", width=11,
        justify="right"),
    "\n")
for(i in 1:nrow(test.mat)) {
  cat(encodeString(test.mat[i,], width=11, justify="right"), aster(Pr.t[i]), "\n")
}

cat("\n", "===Various Statistics in Multiple Regression===", "\n\n")
print(args)

outlier1 <- rep(0,n)
outlier2 <- rep(0,n)
std.sign <- rep(0,n)
for(i in 1:n) {
  outlier1[i] <- ifelse(abs(stdresid[i]) > 2, "out", "in")
  outlier2[i] <- ifelse(abs(stdresid[i]) > 2, 2, 1)
  std.sign[i] <- ifelse(stdresid[i] > 0, 24, 20)
}
abs.stdresid <- abs(stdresid)
ex1 <- as.data.frame(cbind(number=(1:n), stdresid=stdresid,
    pred=pred, y=y,
    outlier1=outlier1, outlier2=outlier2,
    std.sign=std.sign, abs.stdresid=abs.stdresid))
ex1_1 <- as.matrix(ex1[,1:4,8])

```

```

ex1_1 <- as.numeric(ex1_1)
ex1_1 <- matrix(ex1_1,nrow=n,ncol=5)
colnames(ex1_1) <- c("number","stdresid","pred","y","abs.stdresid")
ex2 <- as.data.frame(ex1_1)
ex2 <- ex2[abs.stdresid > 2,]

pred1 <- as.vector(pred) ; abs.stdresid1 <- as.vector(abs.stdresid)
ord <- order(pred1,abs.stdresid1)
re1 <- as.data.frame(t(rbind(pred1,abs.stdresid1)[,ord]))

KDE <- function(x) {
  w <- rep(0,n)
  for (i in 1:n) {
    w[i] <- exp(-(pred[i]-x)^2)/sqrt(2*pi)
  }
  ans <- sum(w*abs(stdresid))/sum(w)
  return(ans)
}
d1 <- seq(min(re1$pred1),max(re1$pred1),0.01)
d2 <- rep(0,length(d1))
for (j in 1:length(d1)) {
  d2[j] <- KDE(d1[j])
}
d3 <- re1$pred1
d4 <- rep(0,length(d3))
for (j in 1:length(d3)) {
  d4[j] <- mean(re1$abs.stdresid1[j:(j+9)],na.rm=TRUE)
}

pairs(data, main="Scatterplot matrix from given data", cex=outlier2,
      pch=21, bg=c("yellow","red")[unclass(ex1$outlier1)])
dev.new()
plot(y~pred, main="Dependent variable VS Fitted values",
     xlab="Fitted response values",
     ylab="Response variables", cex=abs(stdresid)*7, pch=20,
     col=factor(outlier2, labels=c("black", "red")))
abline(0,1,col="blue",lwd=2)
for (i in 1:nrow(ex2)) {
  text(ex2$pred[i], ex2$y[i], ex2$number[i], col="light blue")
}
dev.new()
par(fig=c(0,0.7,0,1), new=FALSE)
plot(pred~stdresid, main="Standard residuals VS predicted values",
     xlab="Studentized residuals", ylab="Fitted response values",
     pch=21, cex=1, col="black")
abline(v=c(-2,0,2),col="saddlebrown",lty=c(2,1,2))
rug(stdresid,side=1,col="lightblue4")
for (i in 1:nrow(ex2)) {
  text(ex2$stdresid[i], ex2$pred[i], ex2$number[i], col="darkgreen", pos=2)
  points(ex2$stdresid[i], ex2$pred[i], pch=19, cex=2, col="tomato")
}

```

```
par(fig=c(0.7,1,0,1),new=TRUE)
boxplot(stdresid,col="green",main="Boxplot of Standard residuals")
for (i in 1:6) {
  text(0.7,summary(stdresid)[i],round(summary(stdresid)[i],3),col="hotpink4")
}
dev.new()
par(fig=c(0,1,0,1),new=FALSE)
plot(abs.stdresid~pred,
      main="Absolute Standard residuals VS predicted values",
      xlab="Fitted response values",
      ylab="Absolute studentized residuals", cex=1.5, pch=std.sign)
legend("topright",c("positive","negative"),pch=c(24,20))
lines(d1,d2,lty=3,lwd=2,col="red")
lines(d3,d4,lty=2,lwd=2,col="blue")
legend("topleft",c("KDE","MA10"),lty=c(3,2),lwd=2,col=c("red","blue"))
dev.new()
qqnorm(stdresid) ; qqline(stdresid)
dev.new()
u <- seq(min(stdresid),max(stdresid),0.001)
hist(stdresid,probability=T,col="bisque",main="Normality of standard residuals")
lines(u,dnorm(u),lwd=2,lty=1,col="darkkhaki")
test.res <- shapiro.test(stdresid)
print(test.res)

return(args)
}

res1 <- MyReg(dat)

str(res1)

attributes(res1)
```

output

```
== ANALYSIS OF VARIANCE ==

  Source      df      SS      MS      F      P-value
-----
Regression    6    195.3    32.55    8.72      0 ***
  Error      93   347.11     3.73
  Total      99   542.41
-----

Estimated error variance : 3.7323
R-squares : 0.3601

== PARAMETER ESTIMATES ==
```

	Estimate	Std.Error	t value	Pr(> t )
(Intercept)	-0.7866	1.182	-0.6655	0.5074
age	0.0198	0.1951	0.1012	0.9196
height	0.5986	0.1953	3.0653	0.0028 **
weight	-0.0803	0.6073	-0.1322	0.8951
smoking	0.2927	0.4105	0.713	0.4776
therapy	1.2801	0.2794	4.5809	0 ***
surgery	1.0944	0.2018	5.4219	0 ***

===Various Statistics in Multiple Regression===

\$Data

	y	age	height	weight	smoking	therapy	surgery
1	7.6185092	-1.207065749	3.4145235	0.300656147	1	2	3
2	11.1395846	0.277429242	2.5252815	0.371353870	0	3	2
3	5.7838752	1.084441177	3.0659935	0.213260667	0	2	1
4	6.2143115	-2.345697703	2.4975222	0.372750095	1	2	1
5	7.7605163	0.429124689	2.1740014	0.248226012	0	2	1
6	7.7740276	0.506055892	3.1669893	0.394101281	0	3	1
7	2.0073677	-0.574739960	2.1037354	0.883828859	0	2	1
8	6.8965796	-0.546631856	3.1681854	0.531965694	1	2	1
9	7.7153296	-0.564451999	3.3549683	0.174454546	0	3	3
10	8.7708591	-0.890037829	2.9478949	0.007488234	1	1	2
11	5.1149679	-0.477192700	2.8040654	0.626116131	0	3	1
12	1.3942546	-0.998386445	2.3509302	1.179623070	1	1	1
13	8.0460598	-0.776253895	1.8902328	0.102581133	1	2	3
14	6.5964155	0.064458817	3.8492742	0.224677931	0	3	1
15	7.1039366	0.959494059	3.0223625	0.563909112	0	3	2
16	1.4979346	-0.110285494	3.8311406	0.167726840	0	2	0
17	3.2428745	-0.511009506	1.7557121	0.201034863	0	3	1
18	6.2127608	-0.911195417	3.1690264	0.254401211	1	3	1
19	8.1941169	-0.837171680	3.6731663	0.693440641	0	2	2
20	6.0718540	2.415835178	2.9737236	0.213521547	1	3	1
21	3.4667472	0.134088220	2.8086078	0.098975684	1	2	1
22	0.4867931	-0.490685897	2.2180934	0.176831684	1	3	1
23	4.7385541	-0.440547872	5.0581620	0.024831943	1	1	0
24	4.5970284	0.459589441	3.7505015	0.250382061	0	2	0
25	7.8639016	-0.693720247	4.8242083	0.757634302	0	3	0
26	6.9645135	-1.448204910	3.0800596	0.147808868	0	3	0
27	5.0848131	0.574755721	2.3685907	0.180896373	0	3	1
28	8.7539894	-1.023655723	1.4867119	0.278347274	0	2	3
29	8.3501663	-0.015138300	2.3639002	0.008344596	1	3	3
30	5.3399869	-0.935948601	3.2263015	0.315793550	0	3	2
31	5.1378487	1.102297546	4.0136903	0.013843684	1	3	1
32	5.0961565	-0.475593079	3.2527501	0.640095477	0	3	1
33	5.0096718	-0.709440038	1.8280517	0.275633695	1	3	0
34	4.4533125	-0.501258061	3.6687143	0.234801962	0	3	0
35	4.9573960	-1.629093469	1.3498991	0.235832515	0	3	1
36	5.3893573	-1.167619262	2.6341478	0.285561219	1	2	1
37	7.7728607	-2.180039649	2.6838817	0.510195492	0	3	2
38	3.9803126	-1.340993192	1.0517540	0.290288645	0	3	1

39	7.9956412	-0.294293859	3.9200575	0.359636780	0	3	0
40	5.5369194	-0.465897540	2.3771284	0.006432923	1	3	0
41	3.2315194	1.449496265	2.6659634	0.770510833	1	3	0
42	6.7268347	-1.068642724	4.3951479	0.226339440	0	2	3
43	3.5572367	-0.855364634	3.6366744	0.403122869	0	1	0
44	3.9056203	-0.280623002	2.8915683	1.045380347	1	1	1
45	5.7657251	-0.994340076	3.5137628	0.115163135	1	3	0
46	6.8175870	-0.968514318	3.3992718	1.304649095	1	2	1
47	6.4152961	-1.107318193	4.6628564	0.274436277	0	2	0
48	4.8616932	-1.251985886	3.2758934	0.175208053	1	3	1
49	5.4079627	-0.523828119	3.5062726	0.239968258	1	2	1
50	4.1239824	-0.496849957	3.3475520	0.032860857	1	2	1
51	5.3127850	-1.806031257	2.6227624	0.465253488	0	3	3
52	1.2245240	-0.582075925	3.0976195	0.999719975	0	2	1
53	6.4538941	-1.108889624	4.6387446	0.048526795	0	2	2
54	7.2596023	-1.014962009	2.1244075	0.016168668	0	3	2
55	7.5936876	-0.162309524	3.1217600	0.165215369	1	3	1
56	1.5583003	0.563055819	4.3621307	0.468037451	0	1	1
57	3.0788377	1.647817473	2.7653789	1.170797202	0	2	1
58	0.4581519	-0.773353424	1.9466172	0.527473505	1	1	0
59	3.3099758	1.605909629	2.1302164	0.523417265	0	2	1
60	5.2972146	-1.157808548	2.6098730	0.240719133	1	2	1
61	7.4377150	0.656588464	2.1526499	0.093950833	1	2	2
62	5.4576686	2.548991071	2.7393606	0.136985760	1	1	3
63	-1.9376925	-0.034760390	2.5855803	0.939535162	0	1	1
64	9.2465962	-0.669633580	2.8169492	0.243493845	1	3	2
65	7.3245442	-0.007604756	3.4070561	0.554455368	0	2	2
66	3.3409581	1.777084448	3.6246331	0.229048919	1	2	0
67	5.8706458	-1.138607737	4.6782057	0.290679589	1	3	2
68	6.0322781	1.367827179	2.9313063	0.310287531	1	3	1
69	4.7213034	1.329564791	2.6791601	1.945855520	0	2	0
70	7.5783331	0.336472797	4.4710057	0.127057828	1	1	2
71	7.6815165	0.006892838	4.7043294	0.030359993	0	3	0
72	4.4577125	-0.455468738	3.0432440	0.048792384	1	3	1
73	7.7402535	-0.366523933	2.6673427	1.275292553	0	3	2
74	9.7061666	0.648286568	1.1777646	0.395085718	1	3	2
75	3.7923051	2.070270861	4.4112624	0.694050412	0	3	0
76	2.3613461	-0.153398412	2.1624176	0.287642700	1	1	0
77	3.1786706	-1.390700947	1.8762372	0.082086647	1	3	2
78	5.1042255	-0.723581777	6.0437659	0.760985052	1	1	0
79	5.2019065	0.258261762	3.2350213	0.231048529	1	3	1
80	2.6156518	-0.317059115	2.9667414	0.237070641	0	2	1
81	5.2341128	-0.177789958	0.2677805	0.204189201	1	1	4
82	6.4998278	-0.169994077	2.9002094	0.474560230	1	1	2
83	-0.9153396	-1.372301886	3.9760317	0.489410169	1	1	0
84	5.9297626	-0.173787170	3.4138689	0.083323763	0	3	2
85	2.6109900	0.850232257	3.9123222	0.172416743	1	2	0
86	6.4501599	0.697608712	4.9837322	0.068972247	1	3	2
87	8.4683504	0.549997351	4.1691085	0.569063747	0	3	2
88	4.8777452	-0.402731975	2.4912630	0.430463151	0	3	0
89	7.8100242	-0.191593770	3.7041802	0.301186011	1	2	3

90	5.2398306	-1.194527880	2.8015837	0.138446892	0	2	1
91	5.2926870	-0.053158819	2.4619292	0.537928137	1	2	2
92	2.6914147	0.255196001	0.1442413	0.048790497	0	3	3
93	7.0006510	1.705964007	2.2103531	0.777457371	0	3	0
94	5.5954422	1.001513252	3.4878146	0.102332856	0	2	2
95	7.2981776	-0.495583443	5.1680325	0.767098073	0	2	3
96	4.8584337	0.355550297	3.5006946	0.120016478	1	1	1
97	7.0215120	-1.134608044	3.6202102	0.780808501	1	3	0
98	8.6139680	0.878203627	2.0340968	0.287524310	0	3	3
99	7.4413886	0.972916753	3.1626547	0.545354279	1	3	0
100	3.9999188	2.121117105	0.9217625	0.191202086	0	3	0
\$Beta							
		[,1]					
		-0.78660499					
age		0.01975009					
height		0.59862082					
weight		-0.08026628					
smoking		0.29271521					
therapy		1.28005300					
surgery		1.09439426					
\$Sigma2							
		[,1]					
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\$Predict							
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[1,]		7.345432					
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[3,]		4.707563					
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[16,]		4.051261					
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[18,]		6.299293					
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[21,]		4.836605					
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[24.]	4.007609
[25.]	5.866912
[26.]	4.856876
[27.]	5.562668
[28.]	5.904101
[29.]	8.043563
[30.]	7.129841
[31.]	6.864001
[32.]	6.034341
[33.]	4.404443
[34.]	5.220976
[35.]	4.904922
[36.]	4.691485
[37.]	6.764963
[38.]	4.727765
[39.]	5.365503
[40.]	4.759550
[41.]	4.908952
[42.]	7.648438
[43.]	2.621186
[44.]	3.522059
[45.]	5.420799
[46.]	5.071638
[47.]	4.520886
[48.]	6.362891
[49.]	5.229931
[50.]	5.152074
[51.]	7.833764
[52.]	4.630455
[53.]	6.713343
[54.]	6.492714
[55.]	6.292947
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[57.]	4.461878
[58.]	1.893837
[59.]	4.132791
[60.]	4.680746
[61.]	5.549052
[62.]	5.748532
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[64.]	7.188573
[65.]	5.957170
[66.]	4.252710
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[97.] 5.428321  
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\$Residuals

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[1.] 0.27307748  
[2.] 4.40988390  
[3.] 1.07631221  
[4.] 1.63487922  
[5.] 3.60266744  
[6.] 1.75189198  
[7.] -2.03757453  
[8.] 1.89292234  
[9.] -0.60461022  
[10.] 4.04941549  
[11.] -0.65187166  
[12.] -1.77921650  
[13.] 1.58869294  
[14.] 0.16097255  
[15.] 0.07865770  
[16.] -2.55332600  
[17.] -1.92985078  
[18.] -0.08653188  
[19.] 2.10518761  
[20.] -0.17951677  
[21.] -1.36985822  
[22.] -5.25778252



[23.]	0.93516377
[24.]	0.58941938
[25.]	1.99698954
[26.]	2.10763788
[27.]	-0.47785447
[28.]	2.84988813
[29.]	0.30660316
[30.]	-1.78985430
[31.]	-1.72615274
[32.]	-0.93818463
[33.]	0.60522834
[34.]	-0.76766374
[35.]	0.05247420
[36.]	0.69787267
[37.]	1.00789823
[38.]	-0.74745238
[39.]	2.63013818
[40.]	0.77736946
[41.]	-1.67743265
[42.]	-0.92160296
[43.]	0.93605032
[44.]	0.38356091
[45.]	0.34492631
[46.]	1.74594924
[47.]	1.89440971
[48.]	-1.50119816
[49.]	0.17803138
[50.]	-1.02809195
[51.]	-2.52097847
[52.]	-3.40593097
[53.]	-0.25944882
[54.]	0.76688853
[55.]	1.30074046
[56.]	-2.61435704
[57.]	-1.38304001
[58.]	-1.43568476
[59.]	-0.82281543
[60.]	0.61646826
[61.]	1.88866260
[62.]	-0.29086317
[63.]	-4.99721750
[64.]	2.05802371
[65.]	1.36737425
[66.]	-0.91175167
[67.]	-2.41906393
[68.]	-0.16523549
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[73.]	1.01078597

[74.] 3.48498284  
[75.] -1.88710167  
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[77.] -3.44548652  
[78.] 0.77551044  
[79.] -1.16186342  
[80.] -2.00290589  
[81.] -0.07002558  
[82.] 1.83019893  
[83.] -4.01525198  
[84.] -1.34607257  
[85.] -1.80017661  
[86.] -2.07650540  
[87.] 0.76510680  
[88.] 0.37537499  
[89.] 0.27118489  
[90.] 0.72955358  
[91.] -0.39185245  
[92.] -3.73279184  
[93.] 2.65264428  
[94.] -0.46629188  
[95.] -0.78083813  
[96.] 0.88489860  
[97.] 1.59319079  
[98.] 1.06531237  
[99.] 2.22644678  
[100.] 0.36803345

\$Standard.Residuals

[1] 0.14621932 2.32114970 0.57153929 0.87974071 1.90940601 0.92011018 -1.08413684  
[8] 0.99306711 -0.32351339 2.17266615 -0.34288737 -0.98076112 0.84919099 0.08485406  
[15] 0.04173664 -1.36122109 -1.02336295 -0.04573283 1.11549389 -0.09804247 -0.71981894  
[22] -2.77953994 0.51186661 0.31366989 1.07755297 1.13136116 -0.25226494 1.53626742  
[29] 0.16453837 -0.94461232 -0.92611157 -0.49376898 0.32502459 -0.40735099 0.02834328  
[36] 0.36761431 0.54242613 -0.40399631 1.39370414 0.41608024 -0.91057152 -0.49983195  
[43] 0.50878722 0.20754560 0.18421732 0.96496966 1.01878534 -0.79633830 0.09331282  
[50] -0.54082447 -1.36291393 -1.81089345 -0.13990659 0.40781113 0.68549359 -1.41232270  
[57] -0.75325283 -0.78080575 -0.44049989 0.32484902 1.00074739 -0.16549547 -2.69924865  
[64] 1.08900085 0.72007226 -0.49087686 -1.31213965 -0.08811307 0.87626859 1.04331695  
[71] 0.97856093 -0.94605065 0.55642828 1.87485168 -1.03107986 0.16624905 -1.84199828  
[78] 0.43186179 -0.61288542 -1.05455478 -0.04005917 0.97389414 -2.16611993 -0.71160574  
[85] -0.95776832 -1.13700689 0.40890568 0.19899248 0.14514993 0.38777677 -0.20652954  
[92] -2.06344447 1.43414533 -0.24914333 -0.43376220 0.47286605 0.86279698 0.57223897  
[99] 1.19066657 0.20592839

\$SSR

[1] 195.2996

\$SSE

[1] 347.1072

\$F.Statistic

[1] 8.721064

\$P.value

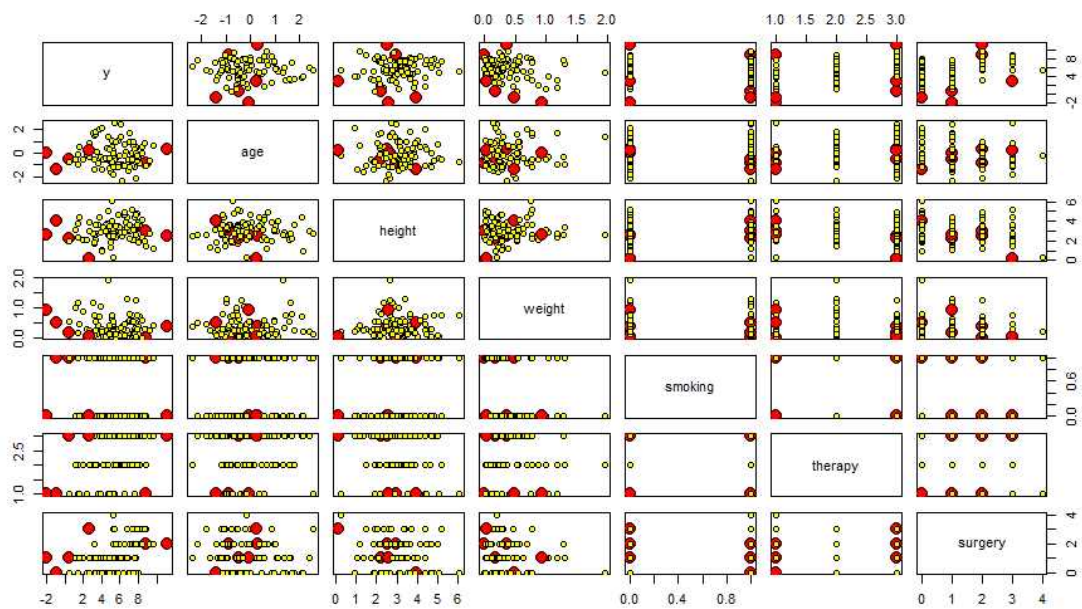
[1] 1.556127e-07

Shapiro-Wilk normality test

data: stdresid

W = 0.9875, p-value = 0.4694

Scatterplot matrix from given data



Dependent variable VS Fitted values

