

COMPUTATIONAL STATISTICS ASSIGNMENT

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Problem 1 : Principal Component Regression

Answers :

BEFORE PCA IN LAST 100 DATA POINTS TEST DATA

```
> before
      y
[1,] -29.58514
[2,] -74.98697
[3,] -57.90908
[4,] -49.40902
[5,] -63.06587
[6,] -95.96224
[7,] -87.91644
[8,] -48.82905
[9,] -63.29259
[10,] -62.49860
[11,] -101.43948
[12,] -55.52751
[13,] -73.71066
[14,] -42.06903
[15,] -88.75302
[16,] -62.42009
[17,] -81.00464
[18,] -67.70099
[19,] -83.79788
[20,] -68.44751
[21,] -56.38498
[22,] -91.29007
[23,] -58.76799
[24,] -38.81030
[25,] -66.56304
[26,] -68.02344
[27,] -50.59879
[28,] -52.78113
[29,] -66.29301
[30,] -67.56520
[31,] -49.61230
[32,] -63.07641
[33,] -57.15544
[34,] -50.69719
```

[35,]	-55.26011
[36,]	-55.31588
[37,]	-55.00962
[38,]	-72.53351
[39,]	-63.51185
[40,]	-63.34615
[41,]	-55.42431
[42,]	-48.78935
[43,]	-72.11494
[44,]	-81.44972
[45,]	-62.19493
[46,]	-82.56678
[47,]	-55.54706
[48,]	-87.42165
[49,]	-70.69117
[50,]	-73.08905
[51,]	-68.53888
[52,]	-61.83007
[53,]	-66.73719
[54,]	-82.77291
[55,]	-49.50351
[56,]	-76.90581
[57,]	-57.13495
[58,]	-55.15680
[59,]	-63.75467
[60,]	-49.23410
[61,]	-84.78451
[62,]	-66.83434
[63,]	-56.34492
[64,]	-41.03712
[65,]	-77.32193
[66,]	-71.11376
[67,]	-79.77172
[68,]	-52.26303
[69,]	-58.03937
[70,]	-70.91842
[71,]	-49.25612
[72,]	-55.68030
[73,]	-60.08965
[74,]	-61.05283
[75,]	-73.36631
[76,]	-92.63440
[77,]	-45.42577
[78,]	-98.09780
[79,]	-77.48607
[80,]	-48.44011
[81,]	-69.59899
[82,]	-61.81468

```
[83,] -81.65876
[84,] -48.46088
[85,] -33.27077
[86,] -85.34737
[87,] -59.02557
[88,] -56.16094
[89,] -64.38483
[90,] -67.71166
[91,] -57.97806
[92,] -67.65507
[93,] -83.06038
[94,] -90.20698
[95,] -67.34685
[96,] -31.55206
[97,] -82.79161
[98,] -81.28438
[99,] -61.83914
[100,] -65.05404
```

AFTER PCA IN LAST 100 DATA POINTS TEST DATA

```
> after
      y
[1,] -30.34441
[2,] -74.50488
[3,] -58.02604
[4,] -49.71689
[5,] -63.21423
[6,] -95.41354
[7,] -86.81427
[8,] -49.11836
[9,] -62.78480
[10,] -62.49982
[11,] -100.59212
[12,] -55.69888
[13,] -73.95451
[14,] -43.07117
[15,] -87.82233
[16,] -62.58515
[17,] -80.33632
[18,] -66.84678
[19,] -83.20964
[20,] -68.54223
[21,] -57.15720
[22,] -90.84295
[23,] -58.73239
[24,] -39.92120
[25,] -66.99185
```

[26,]	-67.28959
[27,]	-50.34699
[28,]	-52.79308
[29,]	-65.30810
[30,]	-67.30659
[31,]	-49.10263
[32,]	-62.91084
[33,]	-57.68099
[34,]	-51.08544
[35,]	-55.77871
[36,]	-55.48919
[37,]	-55.13877
[38,]	-71.98648
[39,]	-63.41935
[40,]	-63.28072
[41,]	-55.12821
[42,]	-49.12467
[43,]	-72.22892
[44,]	-80.87356
[45,]	-62.15756
[46,]	-82.34566
[47,]	-56.17953
[48,]	-86.86995
[49,]	-70.36362
[50,]	-72.61110
[51,]	-68.81426
[52,]	-62.00572
[53,]	-67.20486
[54,]	-82.41622
[55,]	-49.83151
[56,]	-76.72601
[57,]	-57.19517
[58,]	-55.87102
[59,]	-63.87187
[60,]	-49.81014
[61,]	-84.09124
[62,]	-67.09987
[63,]	-56.52025
[64,]	-41.74589
[65,]	-76.26949
[66,]	-70.48629
[67,]	-79.44955
[68,]	-52.35052
[69,]	-58.58133
[70,]	-71.10631
[71,]	-49.61473
[72,]	-55.39126
[73,]	-60.16111

```
[74,] -61.16418
[75,] -73.14781
[76,] -91.40847
[77,] -46.06258
[78,] -96.01739
[79,] -77.33386
[80,] -49.55355
[81,] -69.60571
[82,] -62.26433
[83,] -81.98924
[84,] -49.05402
[85,] -34.25242
[86,] -84.79683
[87,] -59.09573
[88,] -56.25972
[89,] -63.92828
[90,] -67.76024
[91,] -58.53899
[92,] -67.58493
[93,] -82.57345
[94,] -89.60528
[95,] -67.73741
[96,] -32.33692
[97,] -82.44122
[98,] -81.24603
[99,] -62.00686
[100,] -64.98664
```

Graphs: None

Code:

```
library(MASS)
x0 <- matrix(1 , nrow=100, ncol=1 )
x1 <- read.csv("x1train.csv")
x2 <- read.csv("x2train.csv")
x3 <- read.csv("x3train.csv")
x4 <- read.csv("x4train.csv")
y <- read.csv("ytrain.csv")
xx0 <- matrix(1 , nrow=100, ncol=1 )
xx1 <- read.csv("x1test.csv")
xx2 <- read.csv("x2test.csv")
xx3 <- read.csv("x3test.csv")
xx4 <- read.csv("x4test.csv")
```

```

d <- data.frame( x0, x1 , x2 , x3 , x4 )
XX <- data.frame( xx0, xx1 , xx2 , xx3 , xx4 )
XX <- data.matrix(XX)

X=data.matrix(d, rownames.force = NA)
xx <- X
Y=data.matrix(y)

estimator <- ((ginv(t(xx)%*%xx))%*%t(xx))%*%Y

matrix(data=d)
X=data.matrix(d, rownames.force = NA)
dim(Y)
e=(ginv((t(X) %*% X)) %*% t(X) %*% Y)
before<-XX%*%e
print(e)

q<-svd(X)
#We will use the first three parameters as they are explaining 98% of
the variance in the model

rr<-q$v[,c(1,2,3)]
w<-X%*%rr
estimator <- ((ginv(t(w)%*%w))%*%t(w))%*%Y
after<-XX%*%rr%*%estimator

```

Problem 2 : Recursive Least Squares Regression

Answers:

Estimator after each iteration (without forgetting factor)

```

      Y
[1,] -12.5988988
[2,]  0.3496447
[3,] -0.7252545
      Y
[1,] -12.6457218
[2,]  0.3512952
[3,] -0.7286488
      Y

```

```
[1,] -12.6610732
[2,]  0.3507754
[3,] -0.7293641
Y
[1,] -12.6707892
[2,]  0.3505018
[3,] -0.7294149
Y
[1,] -12.6789856
[2,]  0.3513503
[3,] -0.7294277
Y
[1,] -12.6610796
[2,]  0.3514999
[3,] -0.7307829
Y
[1,] -12.6121340
[2,]  0.3489205
[3,] -0.7354518
Y
[1,] -12.6196520
[2,]  0.3480754
[3,] -0.7357627
Y
[1,] -12.6227093
[2,]  0.3482746
[3,] -0.7359187
Y
[1,] -12.6497223
[2,]  0.3510586
[3,] -0.7361123
Y
[1,] -12.6794329
[2,]  0.3519030
[3,] -0.7364692
Y
[1,] -12.6893365
[2,]  0.3533533
[3,] -0.7357555
Y
[1,] -12.6958837
[2,]  0.3530101
[3,] -0.7351384
Y
[1,] -12.6962396
[2,]  0.3530377
[3,] -0.7351074
Y
```

```
[1,] -12.7123377
[2,]  0.3538064
[3,] -0.7339430
Y
[1,] -12.6765379
[2,]  0.3574456
[3,] -0.7323162
Y
[1,] -12.6473669
[2,]  0.3549073
[3,] -0.7317830
Y
[1,] -12.6746031
[2,]  0.3576160
[3,] -0.7304408
Y
[1,] -12.6692821
[2,]  0.3574337
[3,] -0.7299843
Y
[1,] -12.6459771
[2,]  0.3569413
[3,] -0.7315367
Y
[1,] -12.6799366
[2,]  0.3567841
[3,] -0.7274371
Y
[1,] -12.6808347
[2,]  0.3567918
[3,] -0.7274118
Y
[1,] -12.6873175
[2,]  0.3568047
[3,] -0.7279847
Y
[1,] -12.7025625
[2,]  0.3583828
[3,] -0.7259442
Y
[1,] -12.7285107
[2,]  0.3579901
[3,] -0.7279689
Y
[1,] -12.7125439
[2,]  0.3574619
[3,] -0.7266058
Y
```



```
[1,] -12.7059695
[2,]  0.3579124
[3,] -0.7267709
      Y
[1,] -12.7480086
[2,]  0.3585919
[3,] -0.7284715
      Y
[1,] -12.7248720
[2,]  0.3584804
[3,] -0.7310331
      Y
[1,] -12.7733986
[2,]  0.3545170
[3,] -0.7344681
      Y
[1,] -12.8007530
[2,]  0.3565940
[3,] -0.7363903
      Y
[1,] -12.8256380
[2,]  0.3563124
[3,] -0.7370570
      Y
[1,] -12.8095592
[2,]  0.3569197
[3,] -0.7361119
      Y
[1,] -12.7903902
[2,]  0.3571318
[3,] -0.7369658
      Y
[1,] -12.8015307
[2,]  0.3571366
[3,] -0.7365108
      Y
[1,] -12.7875099
[2,]  0.3574493
[3,] -0.7355837
      Y
[1,] -12.8459906
[2,]  0.3609544
[3,] -0.7406004
      Y
[1,] -12.8547818
[2,]  0.3600135
[3,] -0.7413575
      Y
```

```
[1,] -12.8493553
[2,]  0.3597833
[3,] -0.7418741
      Y
[1,] -12.8534587
[2,]  0.3594937
[3,] -0.7421804
      Y
[1,] -12.8509742
[2,]  0.3593926
[3,] -0.7419673
      Y
[1,] -12.8275554
[2,]  0.3576118
[3,] -0.7421610
      Y
[1,] -12.8251863
[2,]  0.3578377
[3,] -0.7421678
      Y
[1,] -12.8240257
[2,]  0.3578801
[3,] -0.7420635
      Y
[1,] -12.8403948
[2,]  0.3581679
[3,] -0.7412919
      Y
[1,] -12.8680845
[2,]  0.3548873
[3,] -0.7409782
      Y
[1,] -12.9038737
[2,]  0.3558674
[3,] -0.7429782
      Y
[1,] -12.8940425
[2,]  0.3553989
[3,] -0.7441960
      Y
[1,] -12.8809675
[2,]  0.3555151
[3,] -0.7453169
      Y
[1,] -12.8763574
[2,]  0.3556935
[3,] -0.7452006
```

Estimator after each iteration forgetting factor = 0.99)

```
Y
[1,] -12.6233565
[2,]  0.3505069
[3,] -0.7270275
Y
[1,] -12.6315113
[2,]  0.3502307
[3,] -0.7274075
Y
[1,] -12.6366384
[2,]  0.3500863
[3,] -0.7274343
Y
[1,] -12.6414011
[2,]  0.3505794
[3,] -0.7274417
Y
[1,] -12.6322456
[2,]  0.3506559
[3,] -0.7281347
Y
[1,] -12.6070178
[2,]  0.3493264
[3,] -0.7305411
Y
[1,] -12.6111171
[2,]  0.3488656
[3,] -0.7307106
Y
[1,] -12.6134405
[2,]  0.3490170
[3,] -0.7308292
Y
[1,] -12.6275642
[2,]  0.3504726
[3,] -0.7309304
Y
[1,] -12.6429473
[2,]  0.3509098
[3,] -0.7311152
Y
[1,] -12.6481547
[2,]  0.3516723
[3,] -0.7307399
Y
```

```
[1,] -12.6509767
[2,]  0.3515244
[3,] -0.7304740
Y
[1,] -12.6511007
[2,]  0.3515341
[3,] -0.7304631
Y
[1,] -12.6592396
[2,]  0.3519227
[3,] -0.7298744
Y
[1,] -12.6408618
[2,]  0.3537908
[3,] -0.7290394
Y
[1,] -12.6268895
[2,]  0.3525751
[3,] -0.7287839
Y
[1,] -12.6410733
[2,]  0.3539857
[3,] -0.7280850
Y
[1,] -12.6396639
[2,]  0.3539374
[3,] -0.7279641
Y
[1,] -12.6278730
[2,]  0.3536882
[3,] -0.7287495
Y
[1,] -12.6449685
[2,]  0.3536091
[3,] -0.7266857
Y
[1,] -12.6456326
[2,]  0.3536148
[3,] -0.7266670
Y
[1,] -12.6496802
[2,]  0.3536229
[3,] -0.7270247
Y
[1,] -12.6577672
[2,]  0.3544600
[3,] -0.7259423
Y
```

```
[1,] -12.6715259
[2,]  0.3542518
[3,] -0.7270158
Y
[1,] -12.6643666
[2,]  0.3540149
[3,] -0.7264046
Y
[1,] -12.6609841
[2,]  0.3542467
[3,] -0.7264896
Y
[1,] -12.6828339
[2,]  0.3545999
[3,] -0.7273735
Y
[1,] -12.6711724
[2,]  0.3545437
[3,] -0.7286646
Y
[1,] -12.6963920
[2,]  0.3524839
[3,] -0.7304498
Y
[1,] -12.7119092
[2,]  0.3536621
[3,] -0.7315402
Y
[1,] -12.7252096
[2,]  0.3535116
[3,] -0.7318965
Y
[1,] -12.7180719
[2,]  0.3537812
[3,] -0.7314769
Y
[1,] -12.708507
[2,]  0.353887
[3,] -0.731903
Y
[1,] -12.7142910
[2,]  0.3538895
[3,] -0.7316668
Y
[1,] -12.7082127
[2,]  0.3540251
[3,] -0.7312649
Y
```

```
[1,] -12.7399527
[2,]  0.3559275
[3,] -0.7339877
Y
[1,] -12.7453494
[2,]  0.3553498
[3,] -0.7344524
Y
[1,] -12.7425993
[2,]  0.3552332
[3,] -0.7347142
Y
[1,] -12.7457802
[2,]  0.3550087
[3,] -0.7349516
Y
[1,] -12.7465162
[2,]  0.3550386
[3,] -0.7350147
Y
[1,] -12.7354842
[2,]  0.3541997
[3,] -0.7351060
Y
[1,] -12.7341484
[2,]  0.3543271
[3,] -0.7351098
Y
[1,] -12.7349450
[2,]  0.3542980
[3,] -0.7351814
Y
[1,] -12.7433509
[2,]  0.3544457
[3,] -0.7347852
Y
[1,] -12.7572668
[2,]  0.3527970
[3,] -0.7346275
Y
[1,] -12.7766310
[2,]  0.3533273
[3,] -0.7357096
Y
[1,] -12.7714512
[2,]  0.3530805
[3,] -0.7363513
Y
```

```
[1,] -12.7646153
[2,]  0.3531412
[3,] -0.7369373
      Y
[1,] -12.7629512
[2,]  0.3532056
[3,] -0.7368953
```

Estimator after each iteration forgetting factor = 0.95)

```
      Y
[1,] -12.8923514
[2,]  0.3568090
[3,] -0.7448459
      Y
[1,] -12.8945319
[2,]  0.3567157
[3,] -0.7449409
      Y
[1,] -12.8964285
[2,]  0.3566478
[3,] -0.7449415
      Y
[1,] -12.8964563
[2,]  0.3566513
[3,] -0.7449414
      Y
[1,] -12.8913611
[2,]  0.3567294
[3,] -0.7453140
      Y
[1,] -12.8770450
[2,]  0.3560858
[3,] -0.7466838
      Y
[1,] -12.8787171
[2,]  0.3558936
[3,] -0.7467396
      Y
[1,] -12.8766735
[2,]  0.3557310
[3,] -0.7466266
      Y
[1,] -12.8812121
[2,]  0.3563007
[3,] -0.7466384
      Y
```

```
[1,] -12.8878322
[2,]  0.3564899
[3,] -0.7466872
Y
[1,] -12.8897409
[2,]  0.3568038
[3,] -0.7465307
Y
[1,] -12.8922619
[2,]  0.3566679
[3,] -0.7463206
Y
[1,] -12.8921215
[2,]  0.3566572
[3,] -0.7463331
Y
[1,] -12.8968382
[2,]  0.3568679
[3,] -0.7459837
Y
[1,] -12.8851684
[2,]  0.3581601
[3,] -0.7455287
Y
[1,] -12.8764498
[2,]  0.3572342
[3,] -0.7454264
Y
[1,] -12.8837869
[2,]  0.3580339
[3,] -0.7450034
Y
[1,] -12.8799077
[2,]  0.3578600
[3,] -0.7446462
Y
[1,] -12.8724251
[2,]  0.3577214
[3,] -0.7451464
Y
[1,] -12.8847369
[2,]  0.3576189
[3,] -0.7438332
Y
[1,] -12.8847394
[2,]  0.3576189
[3,] -0.7438332
Y
```



```
[1,] -12.8836153
[2,]  0.3576154
[3,] -0.7437273
Y
[1,] -12.8891444
[2,]  0.3581497
[3,] -0.7430296
Y
[1,] -12.8939136
[2,]  0.3580626
[3,] -0.7434151
Y
[1,] -12.8872803
[2,]  0.3577918
[3,] -0.7428020
Y
[1,] -12.8851665
[2,]  0.3579411
[3,] -0.7428615
Y
[1,] -12.8961987
[2,]  0.3581379
[3,] -0.7432831
Y
[1,] -12.8886048
[2,]  0.3581312
[3,] -0.7440693
Y
[1,] -12.9018487
[2,]  0.3569386
[3,] -0.7449877
Y
[1,] -12.9081052
[2,]  0.3575120
[3,] -0.7454626
Y
[1,] -12.9152411
[2,]  0.3574181
[3,] -0.7456308
Y
[1,] -12.9094760
[2,]  0.3576589
[3,] -0.7452996
Y
[1,] -12.9030945
[2,]  0.3577465
[3,] -0.7455953
Y
```

```
[1,] -12.9067291
[2,]  0.3577388
[3,] -0.7454384
Y
[1,] -12.9014350
[2,]  0.3578715
[3,] -0.7450899
Y
[1,] -12.9174737
[2,]  0.3589936
[3,] -0.7465787
Y
[1,] -12.9193649
[2,]  0.3587804
[3,] -0.7467379
Y
[1,] -12.9176308
[2,]  0.3587120
[3,] -0.7469027
Y
[1,] -12.9182926
[2,]  0.3586616
[3,] -0.7469517
Y
[1,] -12.9169370
[2,]  0.3586017
[3,] -0.7468274
Y
[1,] -12.9092512
[2,]  0.3579864
[3,] -0.7469116
Y
[1,] -12.9082359
[2,]  0.3580844
[3,] -0.7469163
Y
[1,] -12.9071383
[2,]  0.3581290
[3,] -0.7468159
Y
[1,] -12.9125916
[2,]  0.3582126
[3,] -0.7465493
Y
[1,] -12.9227028
[2,]  0.3570306
[3,] -0.7464266
Y
```

```

[1,] -12.9337580
[2,]  0.3573386
[3,] -0.7470591
      Y
[1,] -12.9304556
[2,]  0.3571929
[3,] -0.7474641
      Y
[1,] -12.9258544
[2,]  0.3572478
[3,] -0.7478523
      Y
[1,] -12.9241199
[2,]  0.3573207
[3,] -0.7478121

```

Estimator after each iteration forgetting factor = 0.2)

```

      Y
[1,] -12.7907345
[2,]  0.3545482
[3,] -0.7375165
      Y
[1,] -12.7959814
[2,]  0.3543287
[3,] -0.7377471
      Y
[1,] -12.7999064
[2,]  0.3541914
[3,] -0.7377505
      Y
[1,] -12.8013657
[2,]  0.3543699
[3,] -0.7377456
      Y
[1,] -12.7921426
[2,]  0.3545039
[3,] -0.7384230
      Y
[1,] -12.7669062
[2,]  0.3533457
[3,] -0.7408394
      Y
[1,] -12.7703346
[2,]  0.3529521
[3,] -0.7409570

```

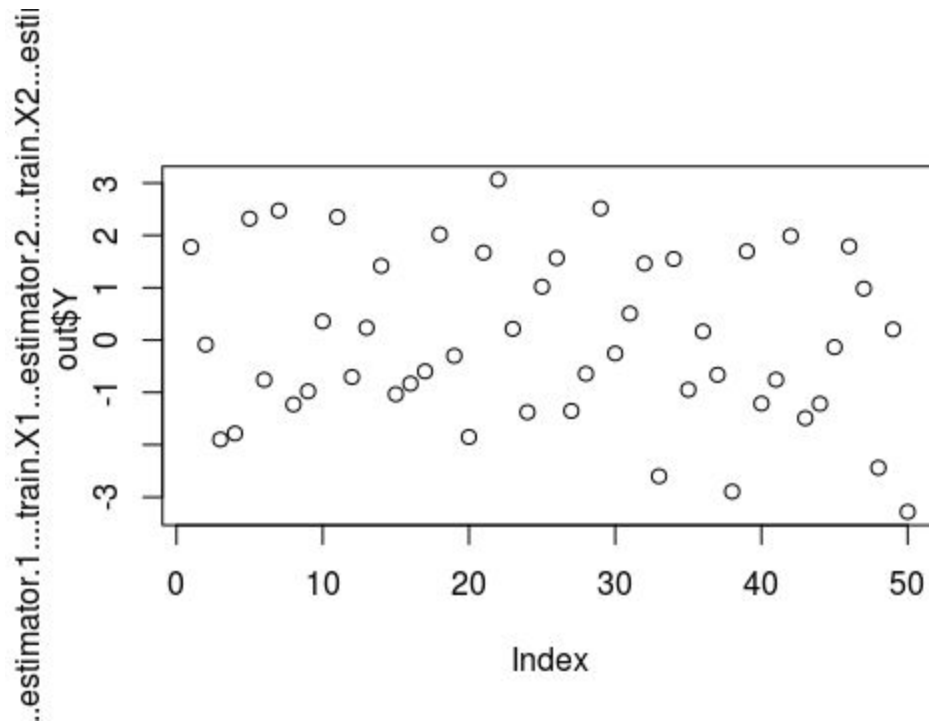
```
Y
[1,] -12.7688314
[2,]  0.3528354
[3,] -0.7408748
Y
[1,] -12.7784950
[2,]  0.3540202
[3,] -0.7409047
Y
[1,] -12.7913071
[2,]  0.3543866
[3,] -0.7410059
Y
[1,] -12.7952541
[2,]  0.3550267
[3,] -0.7406867
Y
[1,] -12.7992356
[2,]  0.3548130
[3,] -0.7403505
Y
[1,] -12.7991390
[2,]  0.3548057
[3,] -0.7403591
Y
[1,] -12.8075250
[2,]  0.3551836
[3,] -0.7397392
Y
[1,] -12.7878665
[2,]  0.3573423
[3,] -0.7389572
Y
[1,] -12.7733843
[2,]  0.3558404
[3,] -0.7387761
Y
[1,] -12.7867386
[2,]  0.3572806
[3,] -0.7380189
Y
[1,] -12.7820267
[2,]  0.3570762
[3,] -0.7375891
Y
[1,] -12.7690096
[2,]  0.3568307
[3,] -0.7384592
```

```
Y
[1,] -12.7898732
[2,]  0.3566663
[3,] -0.7362038
Y
[1,] -12.7901034
[2,]  0.3566679
[3,] -0.7361963
Y
[1,] -12.7904938
[2,]  0.3566691
[3,] -0.7362328
Y
[1,] -12.7997996
[2,]  0.3575758
[3,] -0.7350508
Y
[1,] -12.8102750
[2,]  0.3573886
[3,] -0.7358938
Y
[1,] -12.8005845
[2,]  0.3570033
[3,] -0.7350075
Y
[1,] -12.7968226
[2,]  0.3572680
[3,] -0.7351122
Y
[1,] -12.8173896
[2,]  0.3576307
[3,] -0.7359039
Y
[1,] -12.8038632
[2,]  0.3576121
[3,] -0.7373152
Y
[1,] -12.8283688
[2,]  0.3554305
[3,] -0.7390202
Y
[1,] -12.8408854
[2,]  0.3565500
[3,] -0.7399599
Y
[1,] -12.8540011
[2,]  0.3563805
[3,] -0.7402746
```

```
Y
[1,] -12.8448203
[2,]  0.3567593
[3,] -0.7397454
Y
[1,] -12.8337416
[2,]  0.3569075
[3,] -0.7402564
Y
[1,] -12.8401234
[2,]  0.3568962
[3,] -0.7399826
Y
[1,] -12.8319160
[2,]  0.3570990
[3,] -0.7394419
Y
[1,] -12.8614246
[2,]  0.3591217
[3,] -0.7421513
Y
[1,] -12.8657832
[2,]  0.3586334
[3,] -0.7425194
Y
[1,] -12.8626628
[2,]  0.3585091
[3,] -0.7428160
Y
[1,] -12.8646718
[2,]  0.3583576
[3,] -0.7429650
Y
[1,] -12.8632460
[2,]  0.3582953
[3,] -0.7428355
Y
[1,] -12.8501855
[2,]  0.3572567
[3,] -0.7429740
Y
[1,] -12.8485778
[2,]  0.3574117
[3,] -0.7429810
Y
[1,] -12.8476307
[2,]  0.3574497
[3,] -0.7428945
```

```
Y
[1,] -12.8571766
[2,]  0.3575989
[3,] -0.7424299
Y
[1,] -12.8744097
[2,]  0.3555807
[3,] -0.7422227
Y
[1,] -12.8943674
[2,]  0.3561354
[3,] -0.7433610
Y
[1,] -12.8884805
[2,]  0.3558729
[3,] -0.7440840
Y
[1,] -12.8804315
[2,]  0.3559655
[3,] -0.7447646
Y
[1,] -12.8777243
[2,]  0.3560782
[3,] -0.7447010
```

Graphs :



Training Target values

Code :

```
library(MASS)

train <- read.csv("Problem1_Input_Training.csv")
out <- read.csv("Problem1_Output_Training.csv")

x0 <- matrix(1 , nrow=50, ncol=1 )

d <- data.frame(x0,train$X1,train$X2)
df <- d
y <- data.frame(out)

xx <- as.matrix(d)
y <- as.matrix(out)

estimator <- ((ginv(t(xx)%*%xx))%*%t(xx))%*%(y)
```



```

dd <-
data.frame(x0*estimator[1]+train$X1*estimator[2]+train$X2*estimator[3
])
write.csv(dd,"abc.csv")

plot(dd$x0...estimator.1....train.X1...estimator.2....train.X2...esti
mator.3.-out$Y)

forget_factor <- 0.2    #To be added

real_time <- read.csv("Problem1_Input_Test.csv")
real_output <- read.csv("Problem1_Output_Test.csv")
dd <- data.frame(x0,real_time$X1,real_time$X2)
dd <- as.matrix(dd)

Mn <- ginv(t(xx)%*(xx))
i=1
xx <- as.matrix(df)
Mn <- ginv(t(xx)%*(xx))
d <- data.frame(x0,real_time$X1,real_time$X2)
mx <- as.matrix(d[i,])
estimator <- estimator
+ ((Mn-(Mn%*t(mx))%*mx%*Mn)/(1+mx%*Mn%*t(mx))[1])%*t(mx))*(real_o
utput[i,1]-mx%*estimator)[1]
df <- rbind(df,c(1,real_time$X1[i],real_time$X2[i]))
print(estimator)
xn <- rbind(xx[1:50,],dd[1,])

for(i in 2:50)
{
  xx <- as.matrix(df)
  Mn <- ginv(forget_factor*t(xn)%*(xn)+t(xx)%*xx)
  d <- data.frame(x0,real_time$X1,real_time$X2)
  mx <- as.matrix(d[i,])
  estimator <- estimator
+ ((Mn-(Mn%*t(mx))%*mx%*Mn)/(1+mx%*Mn%*t(mx))[1])%*t(mx))*(real_o
utput[i,1]-mx%*estimator)[1]
  df <- rbind(df,c(1,real_time$X1[i],real_time$X2[i]))
  xn <- rbind(xn,dd[i,])
  print(estimator)
}

```

Problem 3 : Recursive Least Squares Regression

Answers:

Estimator after each iteration (without forgetting factor)

```
Y
[1,] -11.9651919
[2,] -0.0168799
[3,] -0.7623284
Y
[1,] -11.9853679
[2,] -0.0161687
[3,] -0.7637910
Y
[1,] -12.01643309
[2,] -0.01722063
[3,] -0.76523832
Y
[1,] -12.03158175
[2,] -0.01764726
[3,] -0.76531766
Y
[1,] -12.07341752
[2,] -0.01331602
[3,] -0.76538288
Y
[1,] -12.06085026
[2,] -0.01321105
[3,] -0.76633401
Y
[1,] -12.0439477
[2,] -0.0141018
[3,] -0.7679463
Y
[1,] -12.01085711
[2,] -0.01038198
[3,] -0.76657807
Y
[1,] -12.027281256
[2,] -0.009311902
[3,] -0.767416095
Y
[1,] -11.98931347
```

```
[2,] -0.01322493
[3,] -0.76714395
      Y
[1,] -11.97517121
[2,] -0.01362686
[3,] -0.76697406
      Y
[1,] -12.00844377
[2,] -0.00875439
[3,] -0.76457624
      Y
[1,] -11.980084501
[2,] -0.007267774
[3,] -0.767249170
      Y
[1,] -11.992210568
[2,] -0.006324388
[3,] -0.766190866
      Y
[1,] -11.989957403
[2,] -0.006431978
[3,] -0.766353838
      Y
[1,] -12.03030527
[2,] -0.01053341
[3,] -0.76818726
      Y
[1,] -12.01541715
[2,] -0.01182888
[3,] -0.76791509
      Y
[1,] -12.01281516
[2,] -0.01208766
[3,] -0.76804331
      Y
[1,] -12.03314827
[2,] -0.01139097
[3,] -0.76978764
      Y
[1,] -12.07625358
[2,] -0.01048018
[3,] -0.76691627
      Y
[1,] -12.05862026
[2,] -0.01039854
[3,] -0.76904497
      Y
[1,] -12.06716287
```

```
[2,] -0.01032503
[3,] -0.76880463
      Y
[1,] -12.05156191
[2,] -0.01035616
[3,] -0.76742614
      Y
[1,] -12.057812984
[2,] -0.009709085
[3,] -0.766589457
      Y
[1,] -12.11855999
[2,] -0.01062846
[3,] -0.77132941
      Y
[1,] -12.138690838
[2,] -0.009962454
[3,] -0.773047934
      Y
[1,] -12.131169593
[2,] -0.009447043
[3,] -0.773236830
      Y
[1,] -12.141804702
[2,] -0.009275144
[3,] -0.773667050
      Y
[1,] -12.11514881
[2,] -0.00940361
[3,] -0.77661824
      Y
[1,] -12.15000301
[2,] -0.01225028
[3,] -0.77908543
      Y
[1,] -12.10111922
[2,] -0.01596186
[3,] -0.77565035
      Y
[1,] -12.09089540
[2,] -0.01584617
[3,] -0.77537646
      Y
[1,] -12.12678753
[2,] -0.01720183
[3,] -0.77748612
      Y
[1,] -12.15193930
```

```
[2,] -0.01748017
[3,] -0.77636575
      Y
[1,] -12.18798523
[2,] -0.01746471
[3,] -0.77489360
      Y
[1,] -12.18448798
[2,] -0.01738672
[3,] -0.77466235
      Y
[1,] -12.20569943
[2,] -0.01611537
[3,] -0.77648198
      Y
[1,] -12.21346137
[2,] -0.01694613
[3,] -0.77715035
      Y
[1,] -12.20909382
[2,] -0.01713136
[3,] -0.77756616
      Y
[1,] -12.23575667
[2,] -0.01901343
[3,] -0.77955635
      Y
[1,] -12.22987842
[2,] -0.01925251
[3,] -0.77905226
      Y
[1,] -12.22494392
[2,] -0.01962773
[3,] -0.77909308
      Y
[1,] -12.20035397
[2,] -0.01728292
[3,] -0.77916317
      Y
[1,] -12.23181529
[2,] -0.01843262
[3,] -0.78199051
      Y
[1,] -12.19367389
[2,] -0.01910312
[3,] -0.78378845
      Y
[1,] -12.1573573
```

```

[2,] -0.0148004
[3,] -0.7841998
      Y
[1,] -12.13797135
[2,] -0.01533131
[3,] -0.78311652
      Y
[1,] -12.15113702
[2,] -0.01470391
[3,] -0.78148560
      Y
[1,] -12.16393432
[2,] -0.01481759
[3,] -0.78038848
      Y
[1,] -12.15603200
[2,] -0.01451178
[3,] -0.78018903

```

Estimator after each iteration forgetting factor = 0.99)

```

      Y
[1,] -12.1386617
[2,] -0.0146673
[3,] -0.7783137
      Y
[1,] -12.14743366
[2,] -0.01501543
[3,] -0.77870668
      Y
[1,] -12.15224734
[2,] -0.01517498
[3,] -0.77871664
      Y
[1,] -12.16420808
[2,] -0.01378084
[3,] -0.77869294
      Y
[1,] -12.16021078
[2,] -0.01372972
[3,] -0.77898927
      Y
[1,] -12.15465222
[2,] -0.01399614
[3,] -0.77952189
      Y
[1,] -12.14121958

```

```
[2,] -0.01245966
[3,] -0.77903371
      Y
[1,] -12.14466372
[2,] -0.01220583
[3,] -0.77921784
      Y
[1,] -12.13153288
[2,] -0.01373546
[3,] -0.77916240
      Y
[1,] -12.12570373
[2,] -0.01390215
[3,] -0.77910963
      Y
[1,] -12.13740970
[2,] -0.01205953
[3,] -0.77819267
      Y
[1,] -12.12641629
[2,] -0.01147442
[3,] -0.77914899
      Y
[1,] -12.13133750
[2,] -0.01109719
[3,] -0.77871411
      Y
[1,] -12.13078045
[2,] -0.01112277
[3,] -0.77875508
      Y
[1,] -12.14412847
[2,] -0.01255966
[3,] -0.77930891
      Y
[1,] -12.1383604
[2,] -0.0131277
[3,] -0.7792271
      Y
[1,] -12.1373696
[2,] -0.0132321
[3,] -0.7792812
      Y
[1,] -12.14206850
[2,] -0.01304211
[3,] -0.77970151
      Y
[1,] -12.15898251
```

```
[2,] -0.01271108
[3,] -0.77857157
      Y
[1,] -12.15232143
[2,] -0.01266508
[3,] -0.77931382
      Y
[1,] -12.15554115
[2,] -0.01264083
[3,] -0.77921259
      Y
[1,] -12.14808928
[2,] -0.01266021
[3,] -0.77852847
      Y
[1,] -12.15117600
[2,] -0.01235404
[3,] -0.77813056
      Y
[1,] -12.17104283
[2,] -0.01269214
[3,] -0.77971440
      Y
[1,] -12.17700244
[2,] -0.01246807
[3,] -0.78024758
      Y
[1,] -12.17404562
[2,] -0.01226156
[3,] -0.78032751
      Y
[1,] -12.17754087
[2,] -0.01220148
[3,] -0.78046414
      Y
[1,] -12.16689433
[2,] -0.01222721
[3,] -0.78159431
      Y
[1,] -12.17920167
[2,] -0.01329571
[3,] -0.78245596
      Y
[1,] -12.1610641
[2,] -0.0148370
[3,] -0.7811238
      Y
[1,] -12.15676820
```



```
[2,] -0.01478368
[3,] -0.78101702
      Y
[1,] -12.16978025
[2,] -0.01530648
[3,] -0.78177200
      Y
[1,] -12.17989548
[2,] -0.01543446
[3,] -0.78131008
      Y
[1,] -12.19436922
[2,] -0.01545011
[3,] -0.78069786
      Y
[1,] -12.1925868
[2,] -0.0154074
[3,] -0.7805803
      Y
[1,] -12.1996744
[2,] -0.0149418
[3,] -0.7812167
      Y
[1,] -12.2025110
[2,] -0.0152553
[3,] -0.7814579
      Y
[1,] -12.20105092
[2,] -0.01531467
[3,] -0.78159679
      Y
[1,] -12.21148916
[2,] -0.01608646
[3,] -0.78237264
      Y
[1,] -12.2088785
[2,] -0.0161979
[3,] -0.7821399
      Y
[1,] -12.20685916
[2,] -0.01635622
[3,] -0.78215983
      Y
[1,] -12.19715596
[2,] -0.01542367
[3,] -0.78219766
      Y
[1,] -12.20939579
```

```

[2,] -0.01590061
[3,] -0.78330978
      Y
[1,] -12.19413551
[2,] -0.01614887
[3,] -0.78404537
      Y
[1,] -12.17902376
[2,] -0.01437262
[3,] -0.78422371
      Y
[1,] -12.1713113
[2,] -0.0145859
[3,] -0.7837867
      Y
[1,] -12.17670731
[2,] -0.01433989
[3,] -0.78312202
      Y
[1,] -12.18209264
[2,] -0.01439732
[3,] -0.78266456
      Y
[1,] -12.17879026
[2,] -0.01426298
[3,] -0.78258512

```

Estimator after each iteration forgetting factor = 0.95)

```

      Y
[1,] -12.17906505
[2,] -0.01425731
[3,] -0.78265316
      Y
[1,] -12.18536570
[2,] -0.01452676
[3,] -0.78292758
      Y
[1,] -12.18899499
[2,] -0.01465668
[3,] -0.78292882
      Y
[1,] -12.19734647
[2,] -0.01361265
[3,] -0.78289595
      Y

```

```
[1,] -12.19431122
[2,] -0.01356615
[3,] -0.78311790
Y
[1,] -12.19006629
[2,] -0.01375699
[3,] -0.78352405
Y
[1,] -12.17908230
[2,] -0.01249454
[3,] -0.78315753
Y
[1,] -12.1811026
[2,] -0.0123338
[3,] -0.7832692
Y
[1,] -12.17122788
[2,] -0.01357338
[3,] -0.78324358
Y
[1,] -12.16653501
[2,] -0.01370752
[3,] -0.78320894
Y
[1,] -12.17547627
[2,] -0.01223748
[3,] -0.78247554
Y
[1,] -12.16650247
[2,] -0.01175382
[3,] -0.78322334
Y
[1,] -12.17053777
[2,] -0.01144798
[3,] -0.78286482
Y
[1,] -12.1701658
[2,] -0.0114646
[3,] -0.7828924
Y
[1,] -12.18055177
[2,] -0.01261462
[3,] -0.78329727
Y
[1,] -12.17597567
[2,] -0.01310059
[3,] -0.78324355
Y
```

```
[1,] -12.17517490
[2,] -0.01318787
[3,] -0.78328971
Y
[1,] -12.17808905
[2,] -0.01305728
[3,] -0.78355807
Y
[1,] -12.19203595
[2,] -0.01279891
[3,] -0.78262573
Y
[1,] -12.18657865
[2,] -0.01275345
[3,] -0.78320781
Y
[1,] -12.18918907
[2,] -0.01273583
[3,] -0.78312090
Y
[1,] -12.18291957
[2,] -0.01275502
[3,] -0.78253038
Y
[1,] -12.18564399
[2,] -0.01249176
[3,] -0.78218661
Y
[1,] -12.20052550
[2,] -0.01276352
[3,] -0.78338958
Y
[1,] -12.20461145
[2,] -0.01259669
[3,] -0.78376727
Y
[1,] -12.20219673
[2,] -0.01242619
[3,] -0.78383532
Y
[1,] -12.20474460
[2,] -0.01238073
[3,] -0.78393268
Y
[1,] -12.19584006
[2,] -0.01238862
[3,] -0.78485452
Y
```

```
[1,] -12.20537816
[2,] -0.01324748
[3,] -0.78551597
Y
[1,] -12.19097783
[2,] -0.01456724
[3,] -0.78442300
Y
[1,] -12.18731452
[2,] -0.01451902
[3,] -0.78433661
Y
[1,] -12.19756004
[2,] -0.01494696
[3,] -0.78492527
Y
[1,] -12.20607072
[2,] -0.01506379
[3,] -0.78453092
Y
[1,] -12.21820423
[2,] -0.01508956
[3,] -0.78400695
Y
[1,] -12.21646665
[2,] -0.01504599
[3,] -0.78389257
Y
[1,] -12.22167102
[2,] -0.01468188
[3,] -0.78437570
Y
[1,] -12.2236795
[2,] -0.0149083
[3,] -0.7845447
Y
[1,] -12.22255748
[2,] -0.01495253
[3,] -0.78465133
Y
[1,] -12.23092503
[2,] -0.01558972
[3,] -0.78527068
Y
[1,] -12.22848690
[2,] -0.01569748
[3,] -0.78504727
Y
```

```

[1,] -12.22674001
[2,] -0.01583734
[3,] -0.78506640
      Y
[1,] -12.21849810
[2,] -0.01504136
[3,] -0.78510438
      Y
[1,] -12.22831422
[2,] -0.01544027
[3,] -0.78600240
      Y
[1,] -12.21548160
[2,] -0.01563697
[3,] -0.78662975
      Y
[1,] -12.20246184
[2,] -0.01411486
[3,] -0.78678769
      Y
[1,] -12.19589250
[2,] -0.01429789
[3,] -0.78641183
      Y
[1,] -12.20061078
[2,] -0.01408981
[3,] -0.78583321
      Y
[1,] -12.20534875
[2,] -0.01414629
[3,] -0.78543350
      Y
[1,] -12.20244287
[2,] -0.01402411
[3,] -0.78536609

```

Estimator after each iteration forgetting factor = 0.2)

```

      Y
[1,] -11.9822218
[2,] -0.0162796
[3,] -0.7635629
      Y
[1,] -12.00832285
[2,] -0.01716343
[3,] -0.76477898

```

```
Y
[1,] -12.02110220
[2,] -0.01752334
[3,] -0.76484590
Y
[1,] -12.05650243
[2,] -0.01385837
[3,] -0.76490109
Y
[1,] -12.0460430
[2,] -0.0137710
[3,] -0.7656927
Y
[1,] -12.03192539
[2,] -0.01451498
[3,] -0.76703937
Y
[1,] -12.00411378
[2,] -0.01138859
[3,] -0.76588938
Y
[1,] -12.01857396
[2,] -0.01044647
[3,] -0.76662720
Y
[1,] -11.98718501
[2,] -0.01368147
[3,] -0.76640221
Y
[1,] -11.97545070
[2,] -0.01401496
[3,] -0.76626124
Y
[1,] -12.003524071
[2,] -0.009903869
[3,] -0.764238107
Y
[1,] -11.979537033
[2,] -0.008646448
[3,] -0.766498944
Y
[1,] -11.989743202
[2,] -0.007852427
[3,] -0.765608200
Y
[1,] -11.987893032
[2,] -0.007940774
[3,] -0.765742022
```

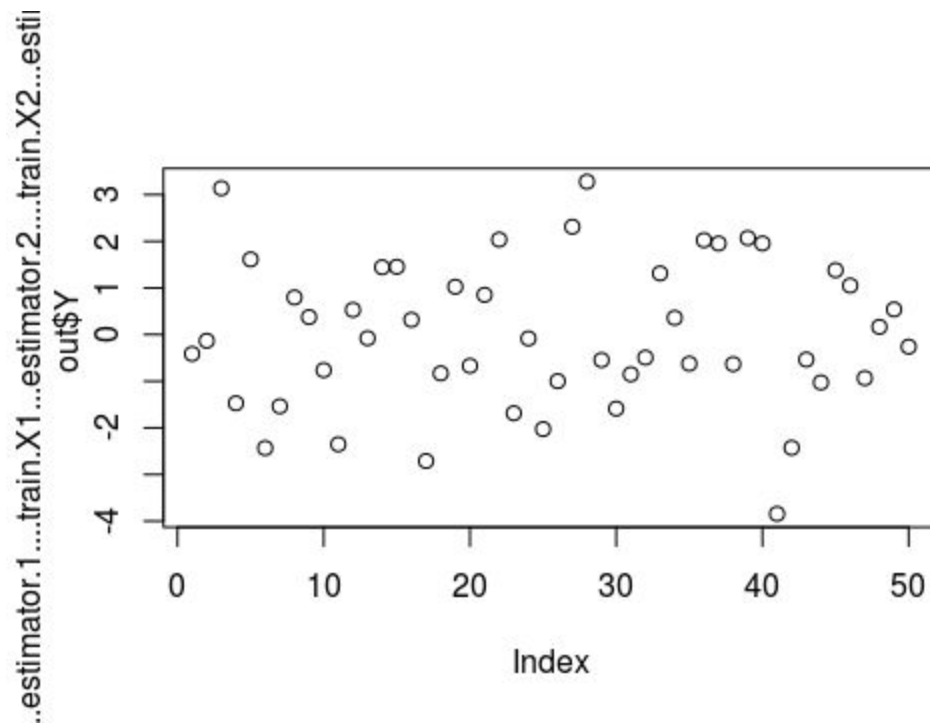
```
Y
[1,] -12.02139765
[2,] -0.01134658
[3,] -0.76726448
Y
[1,] -12.00929342
[2,] -0.01239981
[3,] -0.76704321
Y
[1,] -12.00720325
[2,] -0.01260769
[3,] -0.76714621
Y
[1,] -12.02484669
[2,] -0.01200316
[3,] -0.76865980
Y
[1,] -12.06084915
[2,] -0.01124244
[3,] -0.76626157
Y
[1,] -12.04604509
[2,] -0.01117391
[3,] -0.76804872
Y
[1,] -12.0532386
[2,] -0.0111120
[3,] -0.7678463
Y
[1,] -12.04074975
[2,] -0.01113693
[3,] -0.76674283
Y
[1,] -12.0460468
[2,] -0.0105886
[3,] -0.7660338
Y
[1,] -12.09736787
[2,] -0.01136532
[3,] -0.77003831
Y
[1,] -12.11511712
[2,] -0.01077811
[3,] -0.77155352
Y
[1,] -12.10875990
[2,] -0.01034246
[3,] -0.77171318
```



```
Y
[1,] -12.11813139
[2,] -0.01019099
[3,] -0.77209229
Y
[1,] -12.09569377
[2,] -0.01029913
[3,] -0.77457646
Y
[1,] -12.12544044
[2,] -0.01272864
[3,] -0.77668210
Y
[1,] -12.0852561
[2,] -0.0157797
[3,] -0.7738583
Y
[1,] -12.07696298
[2,] -0.01568586
[3,] -0.77363616
Y
[1,] -12.1074165
[2,] -0.0168361
[3,] -0.7754262
Y
[1,] -12.12840257
[2,] -0.01706834
[3,] -0.77449135
Y
[1,] -12.15855649
[2,] -0.01705541
[3,] -0.77325983
Y
[1,] -12.15623346
[2,] -0.01700361
[3,] -0.77310623
Y
[1,] -12.17466437
[2,] -0.01589892
[3,] -0.77468733
Y
[1,] -12.18201255
[2,] -0.01668539
[3,] -0.77532007
Y
[1,] -12.1783294
[2,] -0.0168416
[3,] -0.7756707
```

```
Y
[1,] -12.20139902
[2,] -0.01847003
[3,] -0.77739270
Y
[1,] -12.19721690
[2,] -0.01864012
[3,] -0.77703407
Y
[1,] -12.19329014
[2,] -0.01893872
[3,] -0.77706655
Y
[1,] -12.17299501
[2,] -0.01700345
[3,] -0.77712440
Y
[1,] -12.20008395
[2,] -0.01799337
[3,] -0.77955881
Y
[1,] -12.16826014
[2,] -0.01855281
[3,] -0.78105894
Y
[1,] -12.13790140
[2,] -0.01495596
[3,] -0.78140284
Y
[1,] -12.12221937
[2,] -0.01538543
[3,] -0.78052651
Y
[1,] -12.13292970
[2,] -0.01487504
[3,] -0.77919975
Y
[1,] -12.14343396
[2,] -0.01496834
[3,] -0.77829921
Y
[1,] -12.13705885
[2,] -0.01472163
[3,] -0.77813831
```

Graphs :



Training Target Variable

Code :

```
library(MASS)

train <- read.csv("Problem1_Input_Training.csv")
out <- read.csv("Problem2_Output_Training.csv")

x0 <- matrix(1 , nrow=50, ncol=1 )

d <- data.frame(x0,train$X1,train$X2)
df <- d
y <- data.frame(out)

xx <- as.matrix(d)
y <- as.matrix(out)

estimator <- ((ginv(t(xx)%*%xx))%*%t(xx))%*%(y)
```

```

dd <-
data.frame(x0*estimator[1]+train$X1*estimator[2]+train$X2*estimator[3
])
write.csv(dd,"abc.csv")

plot(dd$x0...estimator.1....train.X1...estimator.2....train.X2...esti
mator.3.-out$Y)


forget_factor <- 0.2 #To be added

real_time <- read.csv("Problem1_Input_Test.csv")
real_output <- read.csv("Problem2_Output_Test.csv")
dd <- data.frame(x0,real_time$X1,real_time$X2)
dd <- as.matrix(dd)

Mn <- ginv(t(xx)%*(xx))
i=1
xx <- as.matrix(df)
Mn <- ginv(t(xx)%*(xx))
d <- data.frame(x0,real_time$X1,real_time$X2)
mx <- as.matrix(d[i,])
estimator <- estimator
+ ((Mn-(Mn%*t(mx))%*mx%*Mn)/(1+mx%*Mn%*t(mx))[1])%*t(mx))*(real_o
utput[i,1]-mx%*estimator)[1]
df <- rbind(df,c(1,real_time$X1[i],real_time$X2[i]))
print(estimator)
xn <- rbind(xx[1:50,],dd[1,])

for(i in 2:50)
{
  xx <- as.matrix(df)
  Mn <- ginv(forget_factor*t(xn)%*(xn)+t(xx)%*xx)
  d <- data.frame(x0,real_time$X1,real_time$X2)
  mx <- as.matrix(d[i,])
  estimator <- estimator
+ ((Mn-(Mn%*t(mx))%*mx%*Mn)/(1+mx%*Mn%*t(mx))[1])%*t(mx))*(real_o
utput[i,1]-mx%*estimator)[1]
  df <- rbind(df,c(1,real_time$X1[i],real_time$X2[i]))
  xn <- rbind(xn,dd[i,])
  print(estimator)
}

```

Problem 4 : Locally Weighted Regression

Answers:

Each Predicted Value for Heaviside Unit Step Function

[illegible]

Y
0
Y
0
Y
0
Y
0
Y
0
Y
0
Y
0
Y
0

Each Predicted Value for $y = \exp(-di^2/2)$

Y
-5.311127
Y
-34.2156
Y
-14.72224
Y
-2.927352
Y
-25.24385
Y
3.492806
Y
-3.218871
Y
-8.668442
Y
-30.20424
Y
-26.47651
Y
-16.36044
Y
-15.20313
Y
5.478978
Y
-2.091285
Y
-6.922793

	Y
-12.04756	
	Y
-24.03737	
	Y
-11.73891	
	Y
-36.6496	
	Y
-5.057906	
	Y
4.068868	
	Y
-14.02117	
	Y
-30.66851	
	Y
-3.061566	
	Y
-16.49828	
	Y
-36.24089	
	Y
3.661228	
	Y
-19.74716	
	Y
3.176758	
	Y
-16.54649	
	Y
-38.93757	
	Y
-13.97023	
	Y
-21.94231	
	Y
-10.32949	
	Y
-10.61046	
	Y
-27.52245	
	Y
-37.5633	
	Y
-21.9764	
	Y
-3.230242	

	Y
-22.53454	
	Y
-35.13161	
	Y
0.1307241	
	Y
-2.869354	
	Y
-22.27302	
	Y
-6.054649	
	Y
0.1081085	
	Y
-28.59431	
	Y
-18.6208	
	Y
75.4623	
	Y
-15.30552	

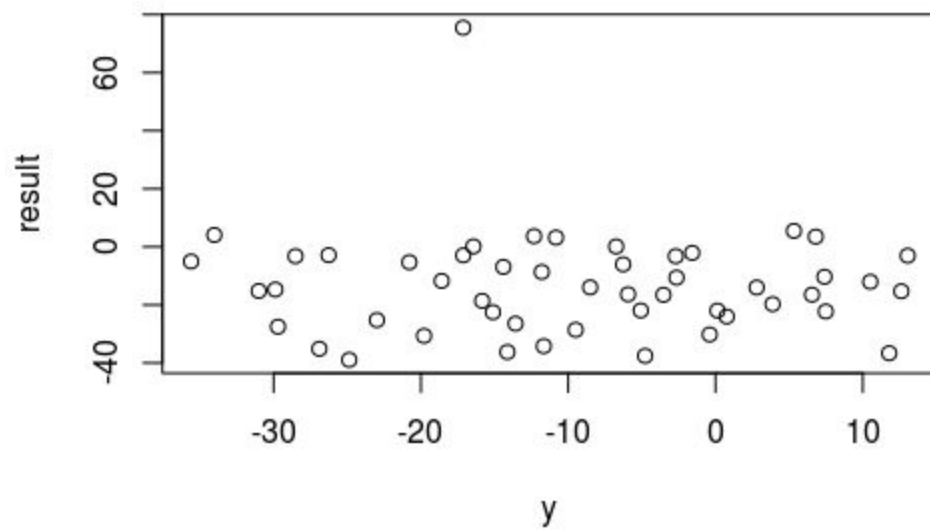
Each Predicted Value for $y = \{ 10 \text{ if } d_i=0 , 1/d_i \text{ otherwise } \}$

	Y
-10.47076	
	Y
-33.31735	
	Y
-15.94794	
	Y
-8.310207	
	Y
-22.3931	
	Y
1.587149	
	Y
-0.7506627	
	Y
-5.513174	
	Y
-30.15696	
	Y
-23.6999	

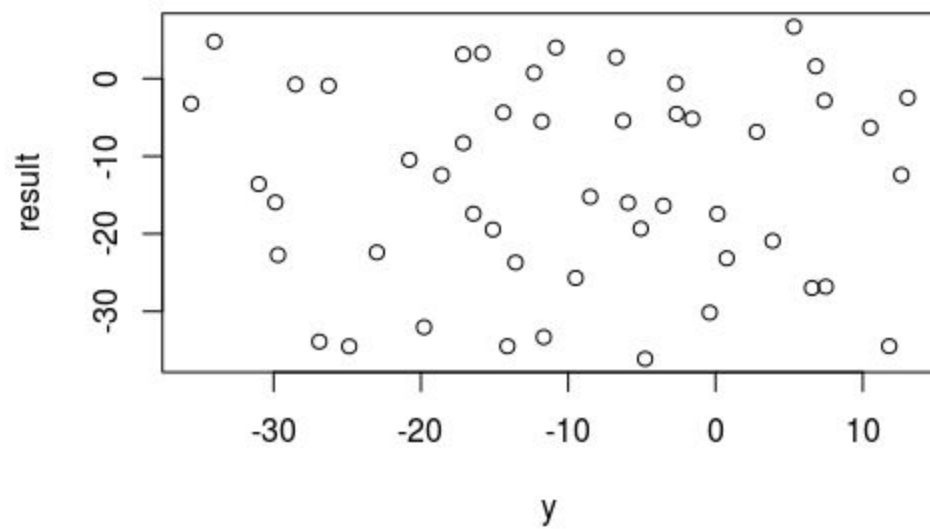
	Y
-16.01439	
	Y
-13.59222	
	Y
6.688057	
	Y
-5.176768	
	Y
-4.371394	
	Y
-6.318124	
	Y
-23.1556	
	Y
-12.43519	
	Y
-34.47836	
	Y
-3.215557	
	Y
4.76911	
	Y
-6.863953	
	Y
-32.03467	
	Y
-2.465819	
	Y
-26.97728	
	Y
-34.49285	
	Y
0.7468434	
	Y
-20.91777	
	Y
4.004062	
	Y
-16.3682	
	Y
-34.5014	
	Y
-15.23251	
	Y
-19.33996	
	Y
-2.853138	

	Y
-4.548527	
	Y
-22.73254	
	Y
-36.10941	
	Y
-17.42033	
	Y
-0.603457	
	Y
-19.4511	
	Y
-33.89007	
	Y
-17.40599	
	Y
-0.9112036	
	Y
-26.81468	
	Y
-5.446286	
	Y
2.741454	
	Y
-25.68753	
	Y
3.281219	
	Y
3.126776	
	Y
-12.41273	

Graphs :



Predicted vs Observed Plot ($y = \exp(-di^2/2)$)



Predicted vs Observed Plot
for $y = \{ 10 \text{ if } di=0, 1/di \text{ otherwise } \}$

Code:

```
train <- read.csv("Problem1_Input_Training.csv")
out <- read.csv("Problem1_Output_Training.csv")
df<- read.csv("Problem1_Input_Training.csv")
real_time_input <- read.csv("Problem1_Input_Test.csv")
y<- read.csv("Problem3_Output_Training.csv")
real_time_output <- read.csv("Problem3_Output_Test.csv")

vi<-diag(50)

x0 <- matrix(1 , nrow=50, ncol=1 )

d <- data.frame(x0,train$X1,train$X2)
df <- d
y <- data.frame(out)

xx <- as.matrix(d)
y <- as.matrix(out)

estimator <- ((ginv(t(xx)%*%xx))%*%t(xx))%*%(y)

dd <-
data.frame(x0*estimator[1]+train$X1*estimator[2]+train$X2*estimator[3
])
write.csv(dd,"abc.csv")

plot(dd$x0...estimator.1....train.X1...estimator.2....train.X2...esti
mator.3.-out$Y) #Shows residual are normally distributed

#Part 1
#Heaviside Unit Step Function
for(i in 1:50)
{
  rr<-diag(50)
  for(j in 1:50)
  {

if(((train$X1[j]-real_time_input$X1[i])^2+(train$X2[j]-real_time_inpu
t$X2[i])^2)<1)
  rr[j,j]=1
```

```

        else
            rr[j,j]=0;
        }
        estimator <-
        ((ginv(t(xx))%*ginv(rr)%*xx))%*t(xx))%*ginv(rr)%*(y)

print(estimator[1,1]*1+estimator[2,1]*real_time_input$X1[i]+estimator
[3,1]*real_time_input$X2[i])
}
#Part 2
#wi=e^-di^2/2
for(i in 1:50)
{
    rr<-diag(50)
    for(j in 1:50)
    {

x<-exp((-((train$X1[j]-real_time_input$X1[i])^2+(train$X2[j]-real_time_
e_input$X2[i])^2))/2)
        rr[j,j]<-x
    }
    estimator <-
    ((ginv(t(xx))%*ginv(rr)%*xx))%*t(xx))%*ginv(rr)%*(y)

print(estimator[1,1]*1+estimator[2,1]*real_time_input$X1[i]+estimator
[3,1]*real_time_input$X2[i])
}
#Part 3
#Inverse Function
for(i in 1:50)
{
    rr<-diag(50)
    for(j in 1:50)
    {

if((train$X1[j]-real_time_input$X1[i])^2+(train$X2[j]-real_time_input
$X2[i])^2==0){
        rr[j,j]=10
    }
    else{

x<-sqrt((train$X1[j]-real_time_input$X1[i])^2+(train$X2[j]-real_time_
input$X2[i])^2)

```

```

        rr[j,j]=1/x;
    }
}
estimator <-
((ginv(t(xx))%*%ginv(rr)%*%xx))%*%t(xx))%*%ginv(rr)%*%(y)

print(estimator[1,1]*1+estimator[2,1]*real_time_input$X1[i]+estimator
[3,1]*real_time_input$X2[i])
}

```