

Q.N. 1) A hospital administrator wished to study the relationship between patients satisfaction (y) and patient's age (x_1 , in years), severity of illness (x_2 , an index) and anxiety level (x_3 , an index). 46 patients are randomly selected and data is collected. The data is as below

ID	y	x_1	x_2	x_3
1	48	50	51	2.3
2	70	41	44	1.8
3	46	42	50	2.2
4	77	29	50	2.1
5	47	38	55	2.2
6	66	36	49	2.0
7	60	33	49	2.1
8	52	44	58	2.9
9	43	47	53	2.5
10	72	32	46	2.6
11	59	33	42	2.0
12	47	40	48	2.2
13	82	29	48	2.5
14	42	47	50	2.6
15	37	44	51	2.6
16	92	28	46	1.8
17	57	36	46	2.3
18	89	28	43	1.8
19	54	45	48	2.4
20	89	29	48	2.4
21	51	34	51	2.3
22	79	33	56	2.5
23	49	55	51	2.4
24	60	43	50	2.3
25	34	55	54	2.5
26	57	32	52	2.4
27	83	36	49	1.8
28	36	53	57	2.8
29	64	30	51	2.4
30	66	43	53	2.3
31	68	45	51	2.2
32	66	40	48	2.2
33	36	49	54	2.9
34	26	52	62	2.9
35	67	43	53	2.4
36	57	53	54	2.2
37	88	29	46	1.9
38	77	29	52	2.3
39	86	23	41	1.8
40	63	25	49	2.0
41	55	42	51	2.7
42	76	31	47	2.0
43	80	34	49	2.2
44	37	47	60	2.4
45	83	22	51	2.0
46	59	37	53	2.1

- a. Fit a multiple linear regression model to the data and state the estimated regression line. How is b_2 interpreted here?

```
> data = scan()  
1: 1 48 50 51 2.3  
6: 2 70 41 44 1.8  
11: 3 46 42 50 2.2  
16: 4 77 29 50 2.1  
21: 5 47 38 55 2.2  
26: 6 66 36 49 2.0  
31: 7 60 33 49 2.1  
36: 8 52 44 58 2.9  
41: 9 43 47 53 2.5  
46: 10 72 32 46 2.6  
51: 11 59 33 42 2.0  
56: 12 47 40 48 2.2  
61: 13 82 29 48 2.5  
66: 14 42 47 50 2.6  
71: 15 37 44 51 2.6  
76: 16 92 28 46 1.8  
81: 17 57 36 46 2.3  
86: 18 89 28 43 1.8  
91: 19 54 45 48 2.4  
96: 20 89 29 48 2.4  
101: 21 51 34 51 2.3  
106: 22 79 33 56 2.5  
111: 23 49 55 51 2.4  
116: 24 60 43 50 2.3  
121: 25 34 55 54 2.5  
126: 26 57 32 52 2.4  
131: 27 83 36 49 1.8  
136: 28 36 53 57 2.8  
141: 29 64 30 51 2.4  
146: 30 66 43 53 2.3  
151: 31 68 45 51 2.2  
156: 32 66 40 48 2.2  
161: 33 36 49 54 2.9  
166: 34 26 52 62 2.9  
171: 35 67 43 53 2.4  
176: 36 57 53 54 2.2  
181: 37 88 29 46 1.9  
186: 38 77 29 52 2.3  
191: 39 86 23 41 1.8  
196: 40 63 25 49 2.0
```

```

201: 41 55 42 51 2.7
206: 42 76 31 47 2.0
211: 43 80 34 49 2.2
216: 44 37 47 60 2.4
221: 45 83 22 51 2.0
226: 46 59 37 53 2.1
231:
Read 230 items
> M = matrix(data, ncol=5, byrow=T)
> M
      [,1][,2][,3][,4][,5]
[1,]  1  48  50  51 2.3
[2,]  2  70  41  44 1.8
[3,]  3  46  42  50 2.2
[4,]  4  77  29  50 2.1
[5,]  5  47  38  55 2.2
[6,]  6  66  36  49 2.0
[7,]  7  60  33  49 2.1
[8,]  8  52  44  58 2.9
[9,]  9  43  47  53 2.5
[10,] 10  72  32  46 2.6
[11,] 11  59  33  42 2.0
[12,] 12  47  40  48 2.2
[13,] 13  82  29  48 2.5
[14,] 14  42  47  50 2.6
[15,] 15  37  44  51 2.6
[16,] 16  92  28  46 1.8
[17,] 17  57  36  46 2.3
[18,] 18  89  28  43 1.8
[19,] 19  54  45  48 2.4
[20,] 20  89  29  48 2.4
[21,] 21  51  34  51 2.3
[22,] 22  79  33  56 2.5
[23,] 23  49  55  51 2.4
[24,] 24  60  43  50 2.3
[25,] 25  34  55  54 2.5
[26,] 26  57  32  52 2.4
[27,] 27  83  36  49 1.8
[28,] 28  36  53  57 2.8
[29,] 29  64  30  51 2.4
[30,] 30  66  43  53 2.3
[31,] 31  68  45  51 2.2
[32,] 32  66  40  48 2.2
[33,] 33  36  49  54 2.9

```

```
[34,] 34 26 52 62 2.9
[35,] 35 67 43 53 2.4
[36,] 36 57 53 54 2.2
[37,] 37 88 29 46 1.9
[38,] 38 77 29 52 2.3
[39,] 39 86 23 41 1.8
[40,] 40 63 25 49 2.0
[41,] 41 55 42 51 2.7
[42,] 42 76 31 47 2.0
[43,] 43 80 34 49 2.2
[44,] 44 37 47 60 2.4
[45,] 45 83 22 51 2.0
[46,] 46 59 37 53 2.1
```

```
> y = M[,2]
```

```
> x1 = M[,3]
```

```
> x2 = M[,3]
```

```
> x1 = M[,3]
```

```
> x2 = M[,4]
```

```
> x3 = M[,5]
```

```
> y
```

```
[1] 48 70 46 77 47 66 60 52 43 72 59 47 82 42 37 92 57 89 54 89 51 79 49 60 34 57 83 36 64 66
68 66 36 26 67 57 88 77 86 63 55 76 80 37 83 59
```

```
> x1
```

```
[1] 50 41 42 29 38 36 33 44 47 32 33 40 29 47 44 28 36 28 45 29 34 33 55 43 55 32 36 53 30 43
45 40 49 52 43 53 29 29 23 25 42 31 34 47 22 37
```

```
> x2
```

```
[1] 51 44 50 50 55 49 49 58 53 46 42 48 48 50 51 46 46 43 48 48 51 56 51 50 54 52 49 57 51 53
51 48 54 62 53 54 46 52 41 49 51 47 49 60 51 53
```

```
> x3
```

```
[1] 2.3 1.8 2.2 2.1 2.2 2.0 2.1 2.9 2.5 2.6 2.0 2.2 2.5 2.6 2.6 1.8 2.3 1.8 2.4 2.4 2.3 2.5 2.4 2.3 2.5
2.4 1.8 2.8 2.4 2.3 2.2 2.2 2.9 2.9 2.4 2.2 1.9 2.3 1.8 2.0 2.7 2.0 2.2 2.4 2.0 2.1
```

```
> model = lm(y~x1+x2+x3)
```

```
> model
```

Call:

```
lm(formula = y ~ x1 + x2 + x3)
```

Coefficients:

(Intercept)	x1	x2	x3
158.491	-1.142	-0.442	-13.470

$Y = 158.491 - 1.142 x_1 - 0.442 x_2 - 13.470 x_3$

b: interpretation –

If x_2 is incremented by 1 unit, then Y is decreased by 0.442

- b. Calculate the coefficient of multiple determination. What does it indicate?

```
> summary(model)
```

Call:

```
lm(formula = y ~ x1 + x2 + x3)
```

Residuals:

Min	1Q	Median	3Q	Max
-18.3524	-6.4230	0.5196	8.3715	17.1601

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	158.4913	18.1259	8.744	5.26e-11 ***
x1	-1.1416	0.2148	-5.315	3.81e-06 ***
x2	-0.4420	0.4920	-0.898	0.3741
x3	-13.4702	7.0997	-1.897	0.0647 .

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 10.06 on 42 degrees of freedom

Multiple R-squared: 0.6822, Adjusted R-squared: 0.6595

F-statistic: 30.05 on 3 and 42 DF, p-value: 1.542e-10

coefficient of multiple determination indicate that 68% of variation in Y can be explained using this model.

- c. Consider the cases 2, 17 and 45. Obtain the DFFITS, DFBETAS and Cook's distance values for these cases to assess their influence.

```
> dffits(model)
```

1	2	3	4	5	6	7	8	9	10	11
12	13	14	15	16	17	18				
0.003365392	0.091451232	-0.246996349	0.050320039	-0.472136428	-0.062609433	-				
0.205259294	0.358895854	-0.101509995	0.281996576	-0.644412890	-0.303009053					
0.456008030	-0.181938302	-0.362555091	0.314681047	-0.250205733	0.194527199					
19	20	21	22	23	24	25	26	27	28	
29	30	31	32	33	34	35	36			
0.012409704	0.568819997	-0.288892901	0.665736971	0.333767209	0.067884177	-				
0.146459074	-0.248559272	0.428074486	0.032517582	-0.151415503	0.222316502					
0.311603405	0.079305897	-0.146607260	-0.346472767	0.243140986	0.554322557					


```

37      38      39      40      41      42      43      44      45      46
0.225642970 0.166353621 -0.162220199 -0.608739745 0.109220429 0.013388989
0.196815026 -0.463926743 -0.040681364 -0.158696762
> cooks.distance(model)
      1      2      3      4      5      6      7      8      9     10     11
12     13     14     15     16     17     18
2.900516e-06 2.139423e-03 1.499893e-02 6.478162e-04 5.412369e-02 1.001951e-03 1.047914e-
02 3.234629e-02 2.624352e-03 2.019402e-02 9.880424e-02 2.222432e-02 5.136521e-02
8.405912e-03 3.208037e-02 2.470201e-02 1.569901e-02 9.608143e-03
      19     20     21     22     23     24     25     26     27     28
29     30     31     32     33     34     35     36
3.943740e-05 7.656783e-02 2.017892e-02 1.051334e-01 2.801513e-02 1.176276e-03 5.468430e-
03 1.545368e-02 4.528550e-02 2.707350e-04 5.828548e-03 1.228245e-02 2.381480e-02
1.604332e-03 5.484993e-03 3.034942e-02 1.448430e-02 7.530661e-02
      37      38      39      40      41      42      43      44      45      46
1.280037e-02 7.030790e-03 6.711882e-03 8.666240e-02 3.046051e-03 4.590508e-05 9.596369e-
03 5.384016e-02 4.237406e-04 6.399528e-03
> dfbetas(model)
      (Intercept)      x1      x2      x3
1  0.0005977634 0.002847889 -0.0006341128 -0.0009826358
2  0.0667690225 0.060203851 -0.0362653798 -0.0465534843
3 -0.0667217871 -0.136382137 0.0176729289 0.0988338741
4 -0.0026372492 -0.032028397 0.0224854005 -0.0123360922
5  0.2494351219 0.090297896 -0.4042663198 0.2723082583
6 -0.0186375583 -0.009511463 -0.0142162462 0.0434781067
7 -0.0510672522 0.059338023 -0.0365873813 0.0676832949
8 -0.2592765845 -0.123105467 0.1081909209 0.1996987750
9  0.0236600857 -0.048291137 0.0037717284 -0.0142146419
10 0.0881603497 -0.090669763 -0.1887171669 0.2448269718
11 -0.5896728730 -0.138109410 0.5151339305 -0.0841365297
12 -0.1812409112 -0.140916164 0.1541563256 0.0171591850
13 0.0626470945 -0.276024733 -0.1768922049 0.3582111028
14 -0.0375319947 -0.075293893 0.1176897412 -0.1080204284
15 -0.0113975594 -0.063121441 0.1793991863 -0.2491336705
16 0.1574109167 -0.053963638 0.0295509954 -0.1787824571
17 -0.1608689658 -0.011828324 0.1990879145 -0.1316542898
18 0.1609428294 -0.002936561 -0.0748397156 -0.0553887483
19 0.0060366064 0.006911791 -0.0091868548 0.0043987924
20 0.0991076408 -0.363089223 -0.1899886913 0.3899851559
21 0.0510228536 0.166479948 -0.0816495185 -0.0380787610
22 -0.4491347867 -0.471110889 0.4432301688 0.0892699630
23 0.0664797118 0.298927532 -0.1129064128 -0.0409802738
24 0.0192973485 0.038208905 -0.0209676397 -0.0038652448
25 0.0206776231 -0.116399385 0.0049904961 0.0222610195

```



```

26 0.0854958784 0.188118427 -0.0787650863 -0.0843506436
27 0.0932095312 0.099089883 0.1573546358 -0.3774349831
28 -0.0182159601 0.010336746 0.0040734580 0.0102249019
29 0.0321428164 0.120999575 -0.0235788056 -0.0689797305
30 -0.0702029968 0.065377757 0.1035080891 -0.0976471158
31 0.0384199516 0.209484367 0.0222142015 -0.1646345482
32 0.0474357874 0.036881679 -0.0403469980 -0.0044910360
33 0.0371805846 -0.016165836 0.0477061728 -0.1117051360
34 0.3044116516 0.012703850 -0.2058902900 -0.0425205531
35 -0.0903192879 0.049090338 0.0754470213 -0.0207646816
36 -0.0650060912 0.407346242 0.1623574060 -0.3626774537
37 0.1315259209 -0.045090781 -0.0141269568 -0.0901781053
38 -0.0569816037 -0.135331586 0.0793796784 0.0190773191
39 -0.1352258938 0.035944346 0.0814164840 0.0023581334
40 -0.0172343213 0.417182739 -0.2498613962 0.1613648361
41 -0.0022583374 -0.007841629 -0.0529451445 0.0923158561
42 0.0072189261 -0.002773363 -0.0010070184 -0.0042619237
43 0.0559647256 -0.063679532 -0.0120363946 0.0043200454
44 0.3366251696 -0.017964869 -0.4044535727 0.2367140455
45 0.0085456656 0.030682507 -0.0247159625 0.0112499915
46 0.0499214671 0.015998781 -0.1190937782 0.1102081200

```

```
> influence.measures(model)
```

```
Influence measures of
```

```
lm(formula = y ~ x1 + x2 + x3):
```

```

      dfb.1_ dfb.x1 dfb.x2 dfb.x3 dffit cov.r cook.d hat inf
1  0.000598 0.00285 -0.000634 -0.000983 0.00337 1.195 2.90e-06 0.0782
2  0.066769 0.06020 -0.036265 -0.046553 0.09145 1.295 2.14e-03 0.1536 *
3 -0.066722 -0.13638 0.017673 0.098834 -0.24700 0.969 1.50e-02 0.0345
4 -0.002637 -0.03203 0.022485 -0.012336 0.05032 1.164 6.48e-04 0.0580
5  0.249435 0.09030 -0.404266 0.272308 -0.47214 0.978 5.41e-02 0.0903
6 -0.018638 -0.00951 -0.014216 0.043478 -0.06261 1.147 1.00e-03 0.0471
7 -0.051067 0.05934 -0.036587 0.067683 -0.20526 1.014 1.05e-02 0.0335
8 -0.259277 -0.12311 0.108191 0.199699 0.35890 1.180 3.23e-02 0.1369
9  0.023660 -0.04829 0.003772 -0.014215 -0.10151 1.126 2.62e-03 0.0434
10 0.088160 -0.09067 -0.188717 0.244827 0.28200 1.308 2.02e-02 0.1860 *
11 -0.589673 -0.13811 0.515134 -0.084137 -0.64441 0.929 9.88e-02 0.1171
12 -0.181241 -0.14092 0.154156 0.017159 -0.30301 0.913 2.22e-02 0.0372
13 0.062647 -0.27602 -0.176892 0.358211 0.45601 1.085 5.14e-02 0.1212
14 -0.037532 -0.07529 0.117690 -0.108020 -0.18194 1.166 8.41e-03 0.0869
15 -0.011398 -0.06312 0.179399 -0.249134 -0.36256 0.967 3.21e-02 0.0610
16 0.157411 -0.05396 0.029551 -0.178782 0.31468 1.081 2.47e-02 0.0832

```


17	-0.160869	-0.01183	0.199088	-0.131654	-0.25021	1.085	1.57e-02	0.0671
18	0.160943	-0.00294	-0.074840	-0.055389	0.19453	1.178	9.61e-03	0.0967
19	0.006037	0.00691	-0.009187	0.004399	0.01241	1.191	3.94e-05	0.0752
20	0.099108	-0.36309	-0.189989	0.389985	0.56882	0.880	7.66e-02	0.0876
21	0.051023	0.16648	-0.081650	-0.038079	-0.28889	0.905	2.02e-02	0.0332
22	-0.449135	-0.47111	0.443230	0.089270	0.66574	0.921	1.05e-01	0.1195
23	0.066480	0.29893	-0.112906	-0.040980	0.33377	1.175	2.80e-02	0.1289
24	0.019297	0.03821	-0.020968	-0.003865	0.06788	1.124	1.18e-03	0.0329
25	0.020678	-0.11640	0.004990	0.022261	-0.14646	1.205	5.47e-03	0.1029
26	0.085496	0.18812	-0.078765	-0.084351	-0.24856	1.065	1.55e-02	0.0594
27	0.093210	0.09909	0.157355	-0.377435	0.42807	1.072	4.53e-02	0.1096
28	-0.018216	0.01034	0.004073	0.010225	0.03252	1.227	2.71e-04	0.1030
29	0.032143	0.12100	-0.023579	-0.068980	-0.15142	1.151	5.83e-03	0.0706
30	-0.070203	0.06538	0.103508	-0.097647	0.22232	1.015	1.23e-02	0.0380
31	0.038420	0.20948	0.022214	-0.164635	0.31160	0.976	2.38e-02	0.0509
32	0.047436	0.03688	-0.040347	-0.004491	0.07931	1.126	1.60e-03	0.0372
33	0.037181	-0.01617	0.047706	-0.111705	-0.14661	1.246	5.48e-03	0.1286
34	0.304412	0.01270	-0.205890	-0.042521	-0.34647	1.282	3.03e-02	0.1843
35	-0.090319	0.04909	0.075447	-0.020765	0.24314	0.952	1.45e-02	0.0309
36	-0.065006	0.40735	0.162357	-0.362677	0.55432	1.078	7.53e-02	0.1429
37	0.131526	-0.04509	-0.014127	-0.090178	0.22564	1.091	1.28e-02	0.0624
38	-0.056982	-0.13533	0.079380	0.019077	0.16635	1.156	7.03e-03	0.0777
39	-0.135226	0.03594	0.081416	0.002358	-0.16222	1.254	6.71e-03	0.1358
40	-0.017234	0.41718	-0.249861	0.161365	-0.60874	0.839	8.67e-02	0.0868
41	-0.002258	-0.00784	-0.052945	0.092316	0.10922	1.196	3.05e-03	0.0900
42	0.007219	-0.00277	-0.001007	-0.004262	0.01339	1.153	4.59e-05	0.0450
43	0.055965	-0.06368	-0.012036	0.004320	0.19682	0.991	9.60e-03	0.0272
44	0.336625	-0.01796	-0.404454	0.236714	-0.46393	1.224	5.38e-02	0.1810
45	0.008546	0.03068	-0.024716	0.011250	-0.04068	1.300	4.24e-04	0.1539 *
46	0.049921	0.01600	-0.119094	0.110208	-0.15870	1.151	6.40e-03	0.0726

If $|DFBETAS| > 2/\sqrt{46}$ i.e 0.3 then its influential

```
> influence.measures(model)$infmtat[c(2,17,45),]
      dfb.1_  dfb.x1  dfb.x2  dfb.x3  dffit cov.r  cook.d  hat
2  0.066769023 0.06020385 -0.03626538 -0.04655348 0.09145123 1.295210 0.0021394233
0.15361084
17 -0.160868966 -0.01182832 0.19908791 -0.13165429 -0.25020573 1.085178 0.0156990120
0.06706793
45 0.008545666 0.03068251 -0.02471596 0.01124999 -0.04068136 1.300270 0.0004237406
0.15385909
```