

Problem statement1:

To record the patient statistics, the agency wants to find the age category of people who frequent the hospital and has the maximum expenditure.

R code

```
5  ***** PROJECT HEALTH CARE *****
6
7  QUESTION 1
8
9  Healthcare <- read.csv("HospitalCosts.csv")
10 View(H Healthcare)
11
12 COUNT_AGE = table(H Healthcare$AGE)
13 COUNT_AGE
14
15 Avg_of_TOTCHG = aggregate(TOTCHG~AGE, Healthcare, mean)
16 Avg_of_TOTCHG
17
18 QUEST1 = cbind(COUNT_AGE, Avg_of_TOTCHG)
19 QUEST1
20
21 hist(hospitalcost$AGE)
22
23 max(COUNT_AGE)
24
25 max(Avg_of_TOTCHG)
```

```
> Healthcare <- read.csv("HospitalCosts.csv")
> view(H Healthcare)
> COUNT_AGE = table(H Healthcare$AGE)
> COUNT_AGE
```

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17
307	10	1	3	2	2	2	3	2	2	4	8	15	18	25	29	29	38	

```
> Avg_of_TOTCHG = aggregate(TOTCHG~AGE, Healthcare, mean)
> Avg_of_TOTCHG
```

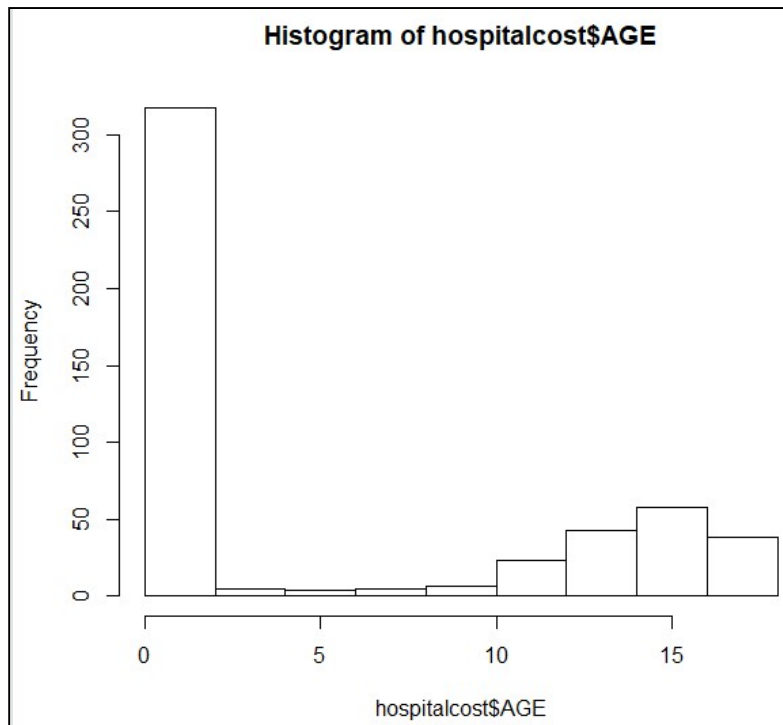
	AGE	TOTCHG
1	0	2208.853
2	1	3774.400
3	2	7298.000
4	3	10183.333
5	4	7996.000
6	5	9253.500
7	6	8964.000
8	7	3362.333
9	8	2370.500
10	9	10573.500
11	10	6117.250
12	11	1781.250
13	12	3660.800
14	13	1729.722
15	14	2585.720
16	15	3853.345
17	16	2384.448
18	17	4599.395

```
>
```

```

> QUEST1 = cbind(COUNT_AGE,Avg_of_TOTCHG)
> QUEST1
  Var1 Freq AGE  TOTCHG
1    0  307  0  2208.853
2    1   10  1  3774.400
3    2    1  2  7298.000
4    3    3  3 10183.333
5    4    2  4  7996.000
6    5    2  5  9253.500
7    6    2  6  8964.000
8    7    3  7  3362.333
9    8    2  8  2370.500
10   9    2  9 10573.500
11  10    4 10  6117.250
12  11    8 11  1781.250
13  12   15 12  3660.800
14  13   18 13  1729.722
15  14   25 14  2585.720
16  15   29 15  3853.345
17  16   29 16  2384.448
18  17   38 17  4599.395
> hist(hospitalcost$AGE)
> hist(hospitalcost$AGE)
> max(COUNT_AGE)
[1] 307
> max(Avg_of_TOTCHG)
[1] 10573.5

```



Based on the output we can see that Age wise Hospital Visit and Expenses

Maximum Hospital Visit – 0-1 yrs age group: 307

Maximum Expenditure – 0-1yrs age group: 10573.5

Problem statement2:

In order of severity of the diagnosis and treatments and to find out the expensive treatments, the agency wants to find the diagnosis related group that has maximum hospitalization and expenditure.

```
27 QUESTION 2
28
29 APRDRG_LOS = aggregate(LOS~APDRG,Healthcare,mean)
30 APRDRG_LOS
31
32 APRDRG_TOTCHG = aggregate(TOTCHG~APDRG,Healthcare,mean)
33 APRDRG_TOTCHG
34
35 QUEST2 = cbind(APRDRG_LOS,APRDRG_TOTCHG)
36 QUEST2
37
38 max(APRDRG_LOS)
39
40 max(APRDRG_TOTCHG)
```

> APRDRG_LOS = aggregate(LOS~APDRG,Healthcare,mean)	31	422	1.6666667
> APRDRG_LOS	32	560	2.0000000
APRDRG LOS	33	561	4.0000000
1 21 2.0000000	34	566	2.0000000
2 23 2.0000000	35	580	1.0000000
3 49 6.0000000	36	581	0.3333333
4 50 2.0000000	37	602	41.0000000
5 51 3.0000000	38	614	15.0000000
6 53 2.9000000	39	626	6.0000000
7 54 1.0000000	40	633	6.2500000
8 57 1.0000000	41	634	4.5000000
9 58 1.0000000	42	636	6.6666667
10 92 1.0000000	43	639	4.5000000
11 97 3.0000000	44	640	2.4419476
12 114 3.0000000	45	710	4.0000000
13 115 4.0000000	46	720	5.0000000
14 137 12.0000000	47	723	1.0000000
15 138 1.5000000	48	740	5.0000000
16 139 1.4000000	49	750	3.0000000
17 141 2.0000000	50	751	2.2857143
18 143 2.0000000	51	753	3.0277778
19 204 3.0000000	52	754	2.2972973
20 206 3.0000000	53	755	1.0000000
21 225 5.5000000	54	756	1.0000000
22 249 1.3333333	55	758	2.6500000
23 254 0.0000000	56	760	7.0000000
24 308 1.0000000	57	776	2.0000000
25 313 1.0000000	58	811	0.5000000
26 317 7.0000000	59	812	1.6666667
27 344 2.5000000	60	863	24.0000000
28 347 0.3333333	61	911	7.0000000
29 420 1.5000000	62	930	5.0000000
30 421 39.0000000	63	952	1.0000000
31 422 1.6666667			

```
> APRDRG_TOTCHG = aggregate(TOTCHG~APRDRG,Healthcare,mean)
> APRDRG_TOTCHG
```

	APRDRG	TOTCHG
1	21	10002.0000
2	23	14174.0000
3	49	20195.0000
4	50	3908.0000
5	51	3023.0000
6	53	8227.1000
7	54	851.0000
8	57	7254.5000
9	58	2117.0000
10	92	12024.0000
11	97	9530.0000
12	114	10562.0000
13	115	12916.0000
14	137	15129.0000
15	138	3405.5000
16	139	3553.2000
17	141	2860.0000
18	143	1393.0000
19	204	8439.0000
20	206	9230.0000
21	225	12824.5000
22	249	2773.6667
23	254	615.0000
24	308	10585.0000
25	313	8159.0000
26	317	17524.0000
27	344	7401.0000
28	347	4199.0000
29	420	3178.5000
30	421	26356.0000
31	422	1725.6667
32	560	2438.5000

28	347	4199.0000
29	420	3178.5000
30	421	26356.0000
31	422	1725.6667
32	560	2438.5000
33	561	2296.0000
34	566	2129.0000
35	580	2825.0000
36	581	2484.3333
37	602	29188.0000
38	614	9177.0000
39	626	3881.5000
40	633	4397.7500
41	634	4976.0000
42	636	7741.3333
43	639	3153.0000
44	640	1640.3670
45	710	8223.0000
46	720	14243.0000
47	723	2644.5000
48	740	11125.0000
49	750	1753.0000
50	751	1547.5714
51	753	2209.5000
52	754	1598.6486
53	755	859.0769
54	756	747.0000
55	758	1747.6500
56	760	4136.5000
57	776	1193.0000
58	811	1919.0000
59	812	3174.6667
60	863	13040.0000
61	911	48388.0000
62	930	13327.0000
63	952	4833.0000


```
> QUEST2 = cbind(APRDRG_LOS, APRDRG_TOTCHG)
> QUEST2
```

	APRDRG	LOS	APRDRG	TOTCHG
1	21	2.0000000	21	10002.0000
2	23	2.0000000	23	14174.0000
3	49	6.0000000	49	20195.0000
4	50	2.0000000	50	3908.0000
5	51	3.0000000	51	3023.0000
6	53	2.9000000	53	8227.1000
7	54	1.0000000	54	851.0000
8	57	1.0000000	57	7254.5000
9	58	1.0000000	58	2117.0000
10	92	1.0000000	92	12024.0000
11	97	3.0000000	97	9530.0000
12	114	3.0000000	114	10562.0000
13	115	4.0000000	115	12916.0000
14	137	12.0000000	137	15129.0000
15	138	1.5000000	138	3405.5000
16	139	1.4000000	139	3553.2000
17	141	2.0000000	141	2860.0000
18	143	2.0000000	143	1393.0000
19	204	3.0000000	204	8439.0000
20	206	3.0000000	206	9230.0000
21	225	5.5000000	225	12824.5000
22	249	1.3333333	249	2773.6667
23	254	0.0000000	254	615.0000
24	308	1.0000000	308	10585.0000
25	313	1.0000000	313	8159.0000
26	317	7.0000000	317	17524.0000
27	344	2.5000000	344	7401.0000
28	347	0.3333333	347	4199.0000
29	420	1.5000000	420	3178.5000
30	421	39.0000000	421	26356.0000

30	421	39.0000000	421	26356.0000
31	422	1.6666667	422	1725.6667
32	560	2.0000000	560	2438.5000
33	561	4.0000000	561	2296.0000
34	566	2.0000000	566	2129.0000
35	580	1.0000000	580	2825.0000
36	581	0.3333333	581	2484.3333
37	602	41.0000000	602	29188.0000
38	614	15.0000000	614	9177.0000
39	626	6.0000000	626	3881.5000
40	633	6.2500000	633	4397.7500
41	634	4.5000000	634	4976.0000
42	636	6.6666667	636	7741.3333
43	639	4.5000000	639	3153.0000
44	640	2.4419476	640	1640.3670
45	710	4.0000000	710	8223.0000
46	720	5.0000000	720	14243.0000
47	723	1.0000000	723	2644.5000
48	740	5.0000000	740	11125.0000
49	750	3.0000000	750	1753.0000
50	751	2.2857143	751	1547.5714
51	753	3.0277778	753	2209.5000
52	754	2.2972973	754	1598.6486
53	755	1.0000000	755	859.0769
54	756	1.0000000	756	747.0000
55	758	2.6500000	758	1747.6500
56	760	7.0000000	760	4136.5000
57	776	2.0000000	776	1193.0000
58	811	0.5000000	811	1919.0000
59	812	1.6666667	812	3174.6667
60	863	24.0000000	863	13040.0000
61	911	7.0000000	911	48388.0000
62	930	5.0000000	930	13327.0000
63	952	1.0000000	952	4833.0000

```
> max(APRDRG_LOS)
[1] 952
> max(APRDRG_TOTCHG)
[1] 48388
> |
```

Based on the output we can see the list of Expenditure based on the Diagnosis and treatment.

TOTCHG - Hospital discharge costs: 48388

APRDRG – All patient Refined Diagnosis Related Groups: 952

Problem statement 3:

To make sure that there is no malpractice, the agency needs to analyze if the race of the patient is related to the hospitalization costs.

```
43 QUESTION 3
44
45 ##### ANOVA
46 Race_mean = aggregate(TOTCHG~RACE,Healthcare,mean)
47 Race_mean
48
49
50 Aov_Hospitalcost = aov(TOTCHG~RACE,Race_mean)
51 summary(Aov_Hospitalcost)
```

```
> Race_mean = aggregate(TOTCHG~RACE,Healthcare,mean)
> Race_mean
  RACE  TOTCHG
1    1 2772.669
2    2 4202.167
3    3 3041.000
4    4 2344.667
5    5 2026.667
6    6 1349.000
> Aov_Hospitalcost = aov(TOTCHG~RACE,Race_mean)
> summary(Aov_Hospitalcost)
              Df Sum Sq Mean Sq F value Pr(>F)
RACE           1 2938135 2938135   6.497 0.0634 .
Residuals      4 1808916   452229
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> |
```

There is no extra charge from one race than another.

Problem statement 4:

To properly utilize the costs, the agency has to analyze the severity of the hospital costs by age and gender for proper allocation of resources.

```
53 QUESTION 4
54
55
56 AGE_GENDER = aggregate(TOTCHG~AGE+FEMALE,Healthcare,mean)
57 AGE_GENDER
```

```
> AGE_GENDER = aggregate(TOTCHG~AGE+FEMALE,Healthcare,mean)
> AGE_GENDER
```

	AGE	FEMALE	TOTCHG
1	0	0	2198.435
2	1	0	4327.750
3	2	0	7298.000
4	3	0	11163.500
5	4	0	9230.000
6	5	0	7923.000
7	6	0	8964.000
8	7	0	3362.333
9	8	0	2370.500
10	9	0	10573.500
11	10	0	7769.667
12	11	0	1468.000
13	12	0	2592.167
14	13	0	1054.000
15	14	0	5741.000
16	15	0	7223.000
17	16	0	4629.833
18	17	0	3961.154
19	0	1	2221.781
20	1	1	1561.000
21	3	1	8223.000
22	4	1	6762.000
23	5	1	10584.000
24	10	1	1160.000
25	11	1	2721.000
26	12	1	4373.222
27	13	1	1922.786
28	14	1	1984.714
29	15	1	2079.842
30	16	1	1798.696
31	17	1	4931.280

Total Charges obtained based on AGE / GENDER (MALR or FEMALE).

Problem statement 5:

Since the length of stay is the crucial factor for inpatients, the agency wants to find if the length of stay can be predicted from age, gender, and race.

```
59 QUESTION 5
60
61 LOS_lm = lm(LOS~AGE+FEMALE+RACE,Healthcare)
62 LOS_lm
63 summary(LOS_lm)
64
65
66 ## Make predictions
67 Predicted_Price = predict(LOS_lm,Healthcare)
68 view(Predicted_Price)
69 Healthcare_Final = cbind(Healthcare,Predicted_Price)
70 view(Healthcare_Final)
71
72
73 install.packages("dplyr")
74
75 Healthcare_Error_pct = transform(Healthcare_Final,Error_pct = (abs(LOS-Predicted_Price)/LOS))
76
77 library(dplyr)
78
79 Healthcare_Error_pct_Final = filter(Healthcare_Error_pct, LOS != 0)
80
81 OverallErrorRate = mean(Healthcare_Error_pct_Final$Error_pct,na.rm=T)
82 OverallErrorRate
83
84 AccuracyRate = 1- OverallErrorRate
85 AccuracyRate
86
87 view(Healthcare_Error_pct_Final)
88
```



```

> LOS_lm = lm(LOS~AGE+FEMALE+RACE,Healthcare)
> LOS_lm

Call:
lm(formula = LOS ~ AGE + FEMALE + RACE, data = Healthcare)

Coefficients:
(Intercept)      AGE      FEMALE      RACE
    2.94377    -0.03960     0.37011    -0.09408

> summary(LOS_lm)

Call:
lm(formula = LOS ~ AGE + FEMALE + RACE, data = Healthcare)

Residuals:
    Min       1Q   Median       3Q      Max
 -3.22  -1.22  -0.85   0.15  37.78

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)   2.94377    0.39318   7.487 3.25e-13 ***
AGE           -0.03960    0.02231  -1.775  0.0766 .
FEMALE         0.37011    0.31024   1.193  0.2334
RACE          -0.09408    0.29312  -0.321  0.7484
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.363 on 495 degrees of freedom
(1 observation deleted due to missingness)
Multiple R-squared:  0.007898, Adjusted R-squared:  0.001886
F-statistic: 1.314 on 3 and 495 DF, p-value: 0.2692

```

For inpatients, Length of stay calculated based on AGE / FEMALE / RACE.

```

> ## Make predictions
> Predicted_Price = predict(LOS_lm,Healthcare)
> view(Predicted_Price)
> Healthcare_Final = cbind(Healthcare,Predicted_Price)
> view(Healthcare_Final)

```

	AGE	FEMALE	LOS	RACE	TOTCHG	APRDRG	Predicted_Price
1	17	1	2	1	2660	560	2.546578
2	17	0	2	1	1689	753	2.176465
3	17	1	7	1	20060	930	2.546578
4	17	1	1	1	736	758	2.546578
5	17	1	1	1	1194	754	2.546578
6	17	0	0	1	3305	347	2.176465
7	17	1	4	1	2205	754	2.546578
8	16	1	2	1	1167	754	2.586180
9	16	1	1	1	532	753	2.586180
10	17	1	2	1	1363	758	2.546578
11	17	1	2	1	1245	758	2.546578
12	15	0	2	1	1656	753	2.255668
13	15	1	2	1	1379	751	2.625782
14	15	1	4	1	2346	758	2.625782
15	15	1	7	1	4006	753	2.625782
16	15	1	4	1	2181	758	2.625782
17	14	1	1	1	628	754	2.665383
18	14	1	4	1	2463	758	2.665383

19	15	1	3	1	1956	753	2.625782
20	14	1	3	1	1802	758	2.665383
21	13	1	1	1	3188	812	2.704985
22	17	1	2	1	2129	566	2.546578
23	12	0	1	1	7421	249	2.374473
24	15	1	1	1	1122	422	2.625782
25	13	1	2	4	1173	754	2.422759
26	12	0	2	1	3625	812	2.374473
27	11	1	2	1	3908	50	2.784188
28	15	0	1	1	3994	139	2.255668
29	11	0	0	1	1033	753	2.414075
30	10	0	2	1	2860	141	2.453677
31	11	0	2	1	3814	420	2.414075
32	7	0	0	1	1132	139	2.572482
33	16	1	2	6	1163	751	2.115803
34	17	1	1	1	610	751	2.546578

```

> library(dplyr)
>
> Healthcare_Error_pct_Final = filter(Healthcare_Error_pct, LOS != 0)
>
> OverallErrorRate = mean(Healthcare_Error_pct_Final$Error_pct,na.rm=T)
> OverallErrorRate
[1] 0.5690605
>
> AccuracyRate = 1- OverallErrorRate
> AccuracyRate
[1] 0.4309395

```

HEALTHCARE PROJECT.R × Healthcare_Error_pct_Final × Predicted_Price × Healthcare_Final × 6.								
Filter								
	AGE	FEMALE	LOS	RACE	TOTCHG	APRDRG	Predicted_Price	Error_pct
1	17	1	2	1	2660	560	2.546578	0.27328916
2	17	0	2	1	1689	753	2.176465	0.08823234
3	17	1	7	1	20060	930	2.546578	0.63620310
4	17	1	1	1	736	758	2.546578	1.54657831
5	17	1	1	1	1194	754	2.546578	1.54657831
6	17	1	4	1	2205	754	2.546578	0.36335542
7	16	1	2	1	1167	754	2.586180	0.29309001
8	16	1	1	1	532	753	2.586180	1.58618001
9	17	1	2	1	1363	758	2.546578	0.27328916
10	17	1	2	1	1245	758	2.546578	0.27328916
11	15	0	2	1	1656	753	2.255668	0.12783404
12	15	1	2	1	1379	751	2.625782	0.31289085
13	15	1	4	1	2346	758	2.625782	0.34355457
14	15	1	7	1	4006	753	2.625782	0.62488833
15	15	1	4	1	2181	758	2.625782	0.34355457
16	14	1	1	1	628	754	2.665383	1.66538341
17	14	1	4	1	2463	758	2.665383	0.33365415

HEALTHCARE PROJECT.R × Healthcare_Error_pct_Final × Predicted_Price × Healthcare_Final × 6.								
Filter								
	AGE	FEMALE	LOS	RACE	TOTCHG	APDRG	Predicted_Price	Error_pct
17	14	1	4	1	2463	758	2.665383	0.33365415
18	15	1	3	1	1956	753	2.625782	0.12473943
19	14	1	3	1	1802	758	2.665383	0.11153886
20	13	1	1	1	3188	812	2.704985	1.70498510
21	17	1	2	1	2129	566	2.546578	0.27328916
22	12	0	1	1	7421	249	2.374473	1.37447317
23	15	1	1	1	1122	422	2.625782	1.62578171
24	13	1	2	4	1173	754	2.422759	0.21137958
25	12	0	2	1	3625	812	2.374473	0.18723659
26	11	1	2	1	3908	50	2.784188	0.39209425
27	15	0	1	1	3994	139	2.255668	1.25566808
28	10	0	2	1	2860	141	2.453677	0.22683828
29	11	0	2	1	3814	420	2.414075	0.20703744
30	16	1	2	6	1163	751	2.115803	0.05790172
31	17	1	1	1	610	751	2.546578	1.54657831
32	6	0	3	1	9530	97	2.612083	0.12930555
33	15	1	1	1	1268	811	2.625782	1.62578171

Problem statement 6:

To perform a complete analysis, the agency wants to find the variable that mainly affects the hospital costs.

```
91 QUESTION 6
92
93 ALL_VAR_lm = lm(TOTCHG~AGE+FEMALE+LOS+RACE+APDRG,Healthcare)
94 ALL_VAR_lm
95 summary(ALL_VAR_lm)
96
97
98 ## Make predictions
99 Predicted_Price = predict(ALL_VAR_lm,Healthcare)
100 view(Predicted_Price)
101
102 Healthcare_Final_All = cbind(Healthcare,Predicted_Price)
103 view(Healthcare_Final_All)
104
105
106 Healthcare_Error_pct_All = transform(Healthcare_Final_All,Error_pct = (abs(TOTCHG-Predicted_Price)/TOTCHG))
107
108 library(dplyr)
109
110 Healthcare_Error_pct_All_Final = filter(Healthcare_Error_pct_All, LOS != 0)
111
112 OverallErrorRate_All = mean(Healthcare_Error_pct_All_Final$Error_pct,na.rm=T)
113 OverallErrorRate_All
114
115 AccuracyRate_All = 1- OverallErrorRate_All
116 AccuracyRate_All
117
118 view(Healthcare_Error_pct_Final)
```

Console

Terminal x

Jobs x

~/Apr Batch/ ↗

```
> ALL_VAR_lm = lm(TOTCHG~AGE+FEMALE+LOS+RACE+APDRG,Healthcare)
> ALL_VAR_lm
```

Call:

```
lm(formula = TOTCHG ~ AGE + FEMALE + LOS + RACE + APRDRG, data = Healthcare)
```

Coefficients:

(Intercept)	AGE	FEMALE	LOS	RACE	APDRG
5218.677	134.695	-390.692	743.152	-212.429	-7.791

```
> summary(ALL_VAR_lm)
```

Call:

```
lm(formula = TOTCHG ~ AGE + FEMALE + LOS + RACE + APRDRG, data = Healthcare)
```

Residuals:

Min	1Q	Median	3Q	Max
-6377	-700	-174	122	43378

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	5218.6769	507.6475	10.280	< 2e-16 ***
AGE	134.6949	17.4711	7.710	7.02e-14 ***
FEMALE	-390.6924	247.7390	-1.577	0.115
LOS	743.1521	34.9225	21.280	< 2e-16 ***
RACE	-212.4291	227.9326	-0.932	0.352
APDRG	-7.7909	0.6816	-11.430	< 2e-16 ***

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

Residual standard error: 2613 on 493 degrees of freedom

(1 observation deleted due to missingness)

Multiple R-squared: 0.5536, Adjusted R-squared: 0.5491

F-statistic: 122.3 on 5 and 493 DF, p-value: < 2.2e-16

```

Console Terminal x Jobs x
~/Apr Batch/ ↗
> ## Make predictions
> Predicted_Price = predict(ALL_VAR_lm,Healthcare)
> View(Predicted_Price)
>
> Healthcare_Final_All = cbind(Healthcare,Predicted_Price)
> View(Healthcare_Final_All)

```

HEALTHCARE PROJECT.R x Predicted_Price x Healthcare_Final_All x							
Filter							
	AGE	FEMALE	LOS	RACE	TOTCHG	APRDRG	Predicted_Price
1	17	1	2	1	2660	560	4028.7415
2	17	0	2	1	1689	753	2915.7810
3	17	1	7	1	20060	930	4861.8511
4	17	1	1	1	736	758	1742.9817
5	17	1	1	1	1194	754	1774.1455
6	17	0	0	1	3305	347	4592.6017
7	17	1	4	1	2205	754	4003.6017
8	16	1	2	1	1167	754	2382.6027
9	16	1	1	1	532	753	1647.2416
10	17	1	2	1	1363	758	2486.1338
11	17	1	2	1	1245	758	2486.1338
12	15	0	2	1	1656	753	2646.3912
13	15	1	2	1	1379	751	2271.2807
14	15	1	4	1	2346	758	3703.0482
15	15	1	7	1	4006	753	5971.4591
16	15	1	4	1	2181	758	3703.0482
17	14	1	1	1	628	754	1370.0609
18	14	1	4	1	2463	758	3568.3533
19	15	1	3	1	1956	753	2998.8508
20	14	1	3	1	1802	758	2825.2012
21	13	1	1	1	3188	812	783.4910

HEALTHCARE PROJECT.R × Predicted_Price × Healthcare_Final_All ×							
Filter							
	AGE	FEMALE	LOS	RACE	TOTCHG	APRDRG	Predicted_Price
21	13	1	1	1	3188	812	783.4910
22	17	1	2	1	2129	566	3981.9958
23	12	0	1	1	7421	249	5425.7924
24	15	1	1	1	1122	422	4091.3505
25	13	1	2	4	1173	754	1341.2309
26	12	0	2	1	3625	812	1782.6406
27	11	1	2	1	3908	50	7193.9558
28	15	0	1	1	3994	139	6686.8813
29	11	0	0	1	1033	753	621.3076
30	10	0	2	1	2860	141	6740.9771
31	11	0	2	1	3814	420	4701.9974
32	7	0	0	1	1132	139	4866.1702
33	16	1	2	1	1163	751	1343.8303
34	17	1	1	1	610	751	1797.5183
35	6	0	3	1	9530	97	7288.1514
36	15	1	1	1	1268	811	1060.6717
37	17	1	4	1	2582	753	4011.3927
38	16	1	2	1	1287	755	2374.8117
39	17	1	3	1	6594	930	1889.2428
40	13	1	0	1	909	755	484.4230
41	7	0	0	1	2530	347	3245.6530

```

> Healthcare_Final_All = cbind(Healthcare,Predicted_Price)
> view(Healthcare_Final_All)
> Healthcare_Error_pct_All = transform(Healthcare_Final_All,Error_pct = (abs(TOTCHG-Predicted_Price)/TOTCHG))
> Healthcare_Error_pct_All_Final = filter(Healthcare_Error_pct_All, LOS != 0)
> overallErrorRate_All = mean(Healthcare_Error_pct_All_Final$Error_pct,na.rm=T)
> overallErrorRate_All
[1] 0.3877528
> AccuracyRate_All = 1- overallErrorRate_All
> AccuracyRate_All
[1] 0.6122472
> view(Healthcare_Error_pct_Final)

```

HEALTHCARE PROJECT.R × Healthcare_Error_pct_Final ×

Filter

	AGE	FEMALE	LOS	RACE	TOTCHG	APRDRG	Predicted_Price	Error_pct
1	17	1	2	1	2660	560	2.546578	0.27328916
2	17	0	2	1	1689	753	2.176465	0.08823234
3	17	1	7	1	20060	930	2.546578	0.63620310
4	17	1	1	1	736	758	2.546578	1.54657831
5	17	1	1	1	1194	754	2.546578	1.54657831
6	17	1	4	1	2205	754	2.546578	0.36335542
7	16	1	2	1	1167	754	2.586180	0.29309001
8	16	1	1	1	532	753	2.586180	1.58618001
9	17	1	2	1	1363	758	2.546578	0.27328916
10	17	1	2	1	1245	758	2.546578	0.27328916
11	15	0	2	1	1656	753	2.255668	0.12783404
12	15	1	2	1	1379	751	2.625782	0.31289085
13	15	1	4	1	2346	758	2.625782	0.34355457
14	15	1	7	1	4006	753	2.625782	0.62488833
15	15	1	4	1	2181	758	2.625782	0.34355457
16	14	1	1	1	628	754	2.665383	1.66538341
17	14	1	4	1	2463	758	2.665383	0.33365415
18	15	1	3	1	1956	753	2.625782	0.12473943
19	14	1	3	1	1802	758	2.665383	0.11153886

HEALTHCARE PROJECT.R × Healthcare_Error_pct_Final ×

Filter

	AGE	FEMALE	LOS	RACE	TOTCHG	APRDRG	Predicted_Price	Error_pct
19	14	1	3	1	1802	758	2.665383	0.11153886
20	13	1	1	1	3188	812	2.704985	1.70498510
21	17	1	2	1	2129	566	2.546578	0.27328916
22	12	0	1	1	7421	249	2.374473	1.37447317
23	15	1	1	1	1122	422	2.625782	1.62578171
24	13	1	2	4	1173	754	2.422759	0.21137958
25	12	0	2	1	3625	812	2.374473	0.18723659
26	11	1	2	1	3908	50	2.784188	0.39209425
27	15	0	1	1	3994	139	2.255668	1.25566808
28	10	0	2	1	2860	141	2.453677	0.22683828
29	11	0	2	1	3814	420	2.414075	0.20703744
30	16	1	2	6	1163	751	2.115803	0.05790172
31	17	1	1	1	610	751	2.546578	1.54657831
32	6	0	3	1	9530	97	2.612083	0.12930555
33	15	1	1	1	1268	811	2.625782	1.62578171
34	17	1	4	1	2582	753	2.546578	0.36335542
35	16	1	2	1	1287	755	2.586180	0.29309001
36	17	1	3	1	6594	930	2.546578	0.15114056
37	11	1	2	2	1534	753	2.690113	0.34505659

HEALTHCARE PROJECT.R × Healthcare_Error_pct_Final ×								
Filter								
	AGE	FEMALE	LOS	RACE	TOTCHG	APRDRG	Predicted_Price	Error_pct
38	3	0	5	1	14243	720	2.730888	0.45382231
39	16	1	3	1	1699	754	2.586180	0.13794000
40	2	0	2	1	7298	53	2.770490	0.38524507
41	16	1	1	1	636	754	2.586180	1.58618001
42	15	1	1	1	626	754	2.625782	1.62578171
43	1	0	2	1	3782	53	2.810092	0.40504592
44	14	1	2	1	1444	753	2.665383	0.33269170
45	14	1	2	1	1183	754	2.665383	0.33269170
46	14	1	5	1	3045	754	2.665383	0.46692332
47	14	1	5	1	3624	754	2.665383	0.46692332
48	14	1	12	1	6810	760	2.665383	0.77788472
49	1	0	1	1	1409	249	2.810092	1.81009185
50	13	0	2	1	1211	754	2.334871	0.16743574
51	1	0	4	1	9606	53	2.810092	0.29747704
52	1	1	1	1	1411	249	3.180205	2.18020548
53	1	0	1	1	2932	249	2.810092	1.81009185
54	1	0	3	1	5075	139	2.810092	0.06330272
55	14	1	1	1	762	753	2.665383	1.66538341
56	16	1	6	1	6329	753	2.586180	0.56897000