

Double-click (or enter) to edit

```
# -----
# R Code for Healthcare Cost Analysis
# -----

# Install necessary packages
install.packages("dplyr")
install.packages("ggplot2")
install.packages("readxl") # <-- ADDED THIS PACKAGE

library(dplyr)
library(ggplot2)
library(readxl) # <-- ADDED THIS LIBRARY

# -----
# 0. Load and Clean Data
# -----

file_path <- "/content/1555054100_hospitalcosts.xlsx"

# --- OLD CODE ---
# df <- read.csv(file_path)
# --- NEW CODE ---
df <- read_excel(file_path) # <-- THIS IS THE FIX

# Clean the data (remove rows with NA values)
df_cleaned <- na.omit(df)

cat("--- Initial Data Structure ---\n")
str(df_cleaned)
```

Installing package into ‘/usr/local/lib/R/site-library’
(as ‘lib’ is unspecified)

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(as ‘lib’ is unspecified)

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(as ‘lib’ is unspecified)

```
--- Initial Data Structure ---
tibble [499 × 6] (S3: tbl_df/tbl/data.frame)
 $ AGE   : num [1:499] 17 17 17 17 17 17 17 16 16 17 ...
 $ FEMALE: num [1:499] 1 0 1 1 1 0 1 1 1 1 ...
 $ LOS   : num [1:499] 2 2 7 1 1 0 4 2 1 2 ...
 $ RACE  : num [1:499] 1 1 1 1 1 1 1 1 1 1 ...
 $ TOTCHG: num [1:499] 2660 1689 20060 736 1194 ...
 $ APRDRG: num [1:499] 560 753 930 758 754 347 754 754 753 758 ...
 - attr(*, "na.action")= 'omit' Named int 277
 ..- attr(*, "names")= chr "277"
```

Double-click (or enter) to edit

```
# -----
# 1. Task 1: Patient Statistics by Age
# -----
cat("\n--- Task 1: Age Category Statistics ---\n")

age_analysis <- df_cleaned %>%
  group_by(AGE) %>%
  summarise(
    TotalExpenditure = sum(TOTCHG),
    VisitCount = n()
  ) %>%
  arrange(desc(VisitCount)) # Sort by visit count

# Print the full analysis
print(age_analysis, n=20) # n=20 to make sure all rows are printed

# Find the age with maximum expenditure
max_exp_age <- age_analysis[which.max(age_analysis$TotalExpenditure), ]
cat("\nAge category with maximum expenditure:\n")
print(max_exp_age)
```

```
# Find the age with most frequent visits
max_visit_age <- age_analysis[which.max(age_analysis$VisitCount), ]
cat("\nAge category with most frequent visits:\n")
print(max_visit_age)
```

```
--- Task 1: Age Category Statistics ---
```

```
# A tibble: 18 × 3
```

	AGE	TotalExpenditure	VisitCount
	<dbl>	<dbl>	<int>
1	0	676962	306
2	17	174777	38
3	15	111747	29
4	16	69149	29
5	14	64643	25
6	13	31135	18
7	12	54912	15
8	1	37744	10
9	11	14250	8
10	10	24469	4
11	3	30550	3
12	7	10087	3
13	4	15992	2
14	5	18507	2
15	6	17928	2
16	8	4741	2
17	9	21147	2
18	2	7298	1

```
Age category with maximum expenditure:
```

```
# A tibble: 1 × 3
```

	AGE	TotalExpenditure	VisitCount
	<dbl>	<dbl>	<int>
1	0	676962	306

```
Age category with most frequent visits:
```

```
# A tibble: 1 × 3
```

	AGE	TotalExpenditure	VisitCount
	<dbl>	<dbl>	<int>
1	0	676962	306

```
# -----
```

```
# 2. Task 2: Diagnosis Group (APDRG) Analysis
```

```
# -----
```

```
cat("\n--- Task 2: Diagnosis Group (APDRG) Statistics ---\n")
```

```
aprdrg_analysis <- df_cleaned %>%
  group_by(APDRG) %>%
  summarise(
    TotalExpenditure = sum(TOTCHG),
    VisitCount = n()
  )
```

```
# Find the APRDRG with maximum expenditure
```

```
max_exp_aprdrg <- aprdrg_analysis[which.max(aprdrg_analysis$TotalExpenditure), ]
cat("\nDiagnosis group (APDRG) with maximum expenditure:\n")
print(max_exp_aprdrg)
```

```
# Find the APRDRG with maximum hospitalization (visits)
```

```
max_visit_aprdrg <- aprdrg_analysis[which.max(aprdrg_analysis$VisitCount), ]
cat("\nDiagnosis group (APDRG) with maximum hospitalization:\n")
print(max_visit_aprdrg)
```

```
--- Task 2: Diagnosis Group (APDRG) Statistics ---
```

```
Diagnosis group (APDRG) with maximum expenditure:
```

```
# A tibble: 1 × 3
```

	APDRG	TotalExpenditure	VisitCount
	<dbl>	<dbl>	<int>
1	640	436822	266

```
Diagnosis group (APDRG) with maximum hospitalization:
```

```
# A tibble: 1 × 3
```

	APDRG	TotalExpenditure	VisitCount
	<dbl>	<dbl>	<int>
1	640	436822	266

```
# -----
```

```
# 3. Task 3: Race and Hospitalization Costs
```

```
# -----
cat("\n--- Task 3: Analysis of Race and Hospitalization Costs (ANOVA) ---\n")

# Convert RACE to a factor (a categorical variable) for ANOVA
# This tells R to treat the race codes (1, 2, 3...) as separate groups
df_cleaned$RACE <- as.factor(df_cleaned$RACE)

# Perform the one-way ANOVA test
# We are testing if TOTCHG changes ~ based on RACE
anova_result <- aov(TOTCHG ~ RACE, data = df_cleaned)

cat("\n--- ANOVA Test Summary ---\n")
print(summary(anova_result))
```

--- Task 3: Analysis of Race and Hospitalization Costs (ANOVA) ---

```
--- ANOVA Test Summary ---
              Df      Sum Sq  Mean Sq F value Pr(>F)
RACE           5 1.859e+07  3718656   0.244  0.943
Residuals    493 7.524e+09 15260687
```

```
# -----
# 4. Task 4: Cost Severity by Age and Gender
# -----
cat("\n--- Task 4: Analysis of Cost Severity by Age and Gender ---\n")

# Calculate mean costs
age_gender_analysis <- df_cleaned %>%
  group_by(AGE, FEMALE) %>%
  summarise(Mean_Cost = mean(TOTCHG), .groups = 'drop')

# Recode FEMALE to 'Male' and 'Female' for the plot
age_gender_analysis$Gender <- ifelse(age_gender_analysis$FEMALE == 0, "Male", "Female")

# Print the table
cat("\n--- Mean Costs by Age and Gender Table ---\n")
print(age_gender_analysis, n = 40) # Print all rows

# Create a grouped bar chart
age_gender_chart <- ggplot(age_gender_analysis, aes(x = factor(AGE), y = Mean_Cost, fill = Gender)) +
  geom_bar(stat = "identity", position = position_dodge()) +
  labs(title = "Mean Hospital Costs by Age and Gender",
       x = "Age",
       y = "Mean Hospital Cost ($)",
       fill = "Gender") +
  theme_minimal()

# Save the chart as a PNG file
ggsave("age_gender_cost_analysis_bar_chart.png", plot = age_gender_chart)

cat("\nChart 'age_gender_cost_analysis_bar_chart.png' has been saved to your Colab files.\n")
```

--- Task 4: Analysis of Cost Severity by Age and Gender ---

--- Mean Costs by Age and Gender Table ---

```
# A tibble: 31 × 4
  AGE FEMALE Mean_Cost Gender
<dbl> <dbl> <dbl> <chr>
1     0     0    2198. Male
2     0     1    2230. Female
3     1     0    4328. Male
4     1     1    1561. Female
5     2     0    7298. Male
6     3     0   11164. Male
7     3     1    8223. Female
8     4     0    9230. Male
9     4     1    6762. Female
10    5     0    7923. Male
11    5     1   10584. Female
12    6     0    8964. Male
13    7     0    3362. Male
14    8     0    2370. Male
15    9     0   10574. Male
16   10     0    7770. Male
17   10     1    1160. Female
18   11     0    1468. Male
19   11     1    2721. Female
```

```

20 12 0 2592. Male
21 12 1 4373. Female
22 13 0 1054 Male
23 13 1 1923. Female
24 14 0 5741 Male
25 14 1 1985. Female
26 15 0 7223 Male
27 15 1 2080. Female
28 16 0 4630. Male
29 16 1 1799. Female
30 17 0 3961. Male
31 17 1 4931. Female
Saving 6.67 x 6.67 in image

```

Chart 'age_gender_cost_analysis_bar_chart.png' has been saved to your Colab files.

```

# -----
# 5. Task 5: Predicting Length of Stay (LOS)
# -----
cat("\n--- Task 5: Predicting Length of Stay (LOS) ---\n")

# Build a linear regression model
# We are testing: LOS ~ AGE + FEMALE + RACE
# We must treat FEMALE and RACE as categorical factors
# Note: df_cleaned$RACE was already converted to a factor in Task 3
los_model <- lm(LOS ~ AGE + as.factor(FEMALE) + RACE, data = df_cleaned)

# Print the model summary
cat("\n--- OLS Regression Model Summary for Predicting LOS ---\n")
print(summary(los_model))

```

```

--- Task 5: Predicting Length of Stay (LOS) ---

--- OLS Regression Model Summary for Predicting LOS ---

Call:
lm(formula = LOS ~ AGE + as.factor(FEMALE) + RACE, data = df_cleaned)

Residuals:
    Min       1Q   Median       3Q      Max
-3.211 -1.211 -0.857   0.143  37.789

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)   2.85687    0.23160   12.335  <2e-16 ***
AGE          -0.03938    0.02258   -1.744   0.0818 .
as.factor(FEMALE)1  0.35391    0.31292    1.131   0.2586
RACE2         -0.37501    1.39568   -0.269   0.7883
RACE3          0.78922    3.38581    0.233   0.8158
RACE4          0.59493    1.95716    0.304   0.7613
RACE5         -0.85687    1.96273   -0.437   0.6626
RACE6         -0.71879    2.39295   -0.300   0.7640
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.376 on 491 degrees of freedom
Multiple R-squared:  0.008699, Adjusted R-squared:  -0.005433
F-statistic: 0.6156 on 7 and 491 DF,  p-value: 0.7432

```

```

# -----
# 6. Task 6: Main Variable Affecting Hospital Costs
# -----
cat("\n--- Task 6: Finding Variable that Mainly Affects Hospital Costs ---\n")

# Build a comprehensive linear regression model
# We are testing: TOTCHG ~ AGE + FEMALE + LOS + RACE + APRDRG
# We must treat FEMALE, RACE, and APRDRG as categorical factors
# Note: df_cleaned$RACE was already a factor from Task 3

cost_model <- lm(TOTCHG ~ AGE + as.factor(FEMALE) + LOS + RACE + as.factor(APRDRG),
  data = df_cleaned)

# Print the model summary
cat("\n--- OLS Regression Model Summary for Predicting Hospital Costs (TOTCHG) ---\n")
print(summary(cost_model))

```

```

--- Task 6: Finding Variable that Mainly Affects Hospital Costs ---

```

```
--- OLS Regression Model Summary for Predicting Hospital Costs (TOTCHG) ---
```

```
Call:
```

```
lm(formula = TOTCHG ~ AGE + as.factor(FEMALE) + LOS + RACE +  
  as.factor(APDRG), data = df_cleaned)
```

```
Residuals:
```

```
      Min       1Q   Median       3Q      Max  
-5403.7 -188.8   -52.0   113.5  5403.7
```

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	7017.4364	966.0317	7.264	1.79e-12 ***
AGE	86.5944	20.7881	4.166	3.76e-05 ***
as.factor(FEMALE)1	-136.8780	78.7821	-1.737	0.083032 .
LOS	664.6593	21.2924	31.216	< 2e-16 ***
RACE2	269.7343	408.6436	0.660	0.509563
RACE3	641.3334	862.2531	0.744	0.457413
RACE4	106.4079	458.4198	0.232	0.816557
RACE5	1577.1875	908.2736	1.736	0.083201 .
RACE6	-73.8266	566.3145	-0.130	0.896340
as.factor(APDRG)23	4355.1399	1182.4224	3.683	0.000260 ***
as.factor(APDRG)49	7890.6917	1187.2479	6.646	9.18e-11 ***
as.factor(APDRG)50	-5254.4156	1194.6819	-4.398	1.38e-05 ***
as.factor(APDRG)51	-7323.6414	1184.2871	-6.184	1.46e-09 ***
as.factor(APDRG)53	-1199.9825	954.2018	-1.258	0.209230
as.factor(APDRG)54	-8166.3229	1184.4591	-6.895	1.95e-11 ***
as.factor(APDRG)57	-860.5678	1081.7666	-0.796	0.426752
as.factor(APDRG)58	-5651.6901	1238.0309	-4.565	6.54e-06 ***
as.factor(APDRG)92	3042.9880	1184.6409	2.569	0.010546 *
as.factor(APDRG)97	-0.9807	1211.2219	-0.001	0.999354
as.factor(APDRG)114	771.2360	1199.1537	0.643	0.520471
as.factor(APDRG)115	2529.0158	1063.9012	2.377	0.017887 *
as.factor(APDRG)137	135.6525	1262.5545	0.107	0.914488
as.factor(APDRG)138	-4574.7058	1042.1335	-4.390	1.43e-05 ***
as.factor(APDRG)139	-4931.6448	985.5923	-5.004	8.23e-07 ***
as.factor(APDRG)141	-6352.6992	1195.6625	-5.313	1.74e-07 ***
as.factor(APDRG)143	-8530.9425	1540.3888	-5.538	5.34e-08 ***
as.factor(APDRG)204	-2044.5193	1182.5513	-1.729	0.084547 .
as.factor(APDRG)206	-127.7919	1220.9720	-0.105	0.916691
as.factor(APDRG)225	895.8186	1049.4715	0.854	0.393810
as.factor(APDRG)249	-5315.2746	997.4554	-5.329	1.60e-07 ***
as.factor(APDRG)254	-7979.6240	1540.5665	-5.180	3.43e-07 ***
as.factor(APDRG)308	2123.5545	1199.1936	1.771	0.077303 .
as.factor(APDRG)313	-1178.3407	1110.6267	-1.061	0.289302
as.factor(APDRG)317	4988.0046	1200.2669	4.156	3.92e-05 ***
as.factor(APDRG)344	-2162.1842	1056.0034	-2.048	0.041217 *
as.factor(APDRG)347	-3802.5781	1012.6388	-3.755	0.000197 ***
as.factor(APDRG)420	-6004.9500	1049.0367	-5.724	1.96e-08 ***
as.factor(APDRG)421	-6583.1473	1473.4757	-4.468	1.01e-05 ***
as.factor(APDRG)422	-7058.7682	1015.5830	-6.950	1.37e-11 ***
as.factor(APDRG)560	-7243.4821	1045.9573	-6.925	1.60e-11 ***
as.factor(APDRG)561	-8455.5174	1188.4307	-7.115	4.75e-12 ***
as.factor(APDRG)566	-7552.9821	1184.1817	-6.378	4.65e-10 ***
as.factor(APDRG)580	-4857.0957	1244.2640	-3.902	0.000110 ***