Summary of the Hospital Emergency Room Visit Analysis

This report analyzes 9,216 emergency room visits from a hospital dataset. The data includes patient demographics, wait times, department referrals, and satisfaction scores. Below are the key insights:

1. Data Overview

- Total Records: 9,216
- Columns: 11 (Date, Patient ID, Gender, Age, Satisfaction Score, Race, Wait Time, Department Referral, etc.)
- Missing Values:
 - Satisfaction Score: 72.7% missing (6,699 entries)
 - Department Referral: 58.6% missing (5,400 entries)

2. Patient Demographics

- Gender Distribution:
 - Male: 4,705
 - Female: 4,487
 - Non-Classified (NC): 24
- Race Breakdown:
 - White: 2,571 (largest group)
 - African American: 1,951
 - Two or More Races: 1,557
 - Asian: 1,060
 - Other categories include: Pacific Islander, Native American, Declined to Identify

3. Emergency Room Visit Trends

• Most Common Department Referrals:

- General Practice: 1,840 visits
- Orthopedics: 995 visits
- Physiotherapy: 276 visits
- Others: Cardiology, Neurology, Gastroenterology, and Renal

Yearly Breakdown:

• 2019:

■ White: 1,225 visits

African American: 920 visits

• 2020:

White: 1,346 visits

African American: 1.031 visits

• Admin Flag (Patient Admitted or Not):

True: 4,612False: 4,604

4. Wait Time & Satisfaction Analysis

- Wait Time Distribution (Grouped by Department & Gender)
 - On average, wait times were higher in Physiotherapy & Cardiology
 - Males and Females had similar wait times across departments
- Satisfaction Score by Category:
 - By Age Group:

• Adults: 5.03

o Children: 5.01

o Seniors: 4.95

By Gender:

o Male: 5.03

o Female: 4.96

Non-Classified (NC): 3.20

■ By Race:

• Highest Satisfaction: Pacific Islander (5.33), Native American (5.12)

Lowest Satisfaction: Two or More Races (4.83), White (4.94)

5. Visualizations & Data Processing

- Bar Charts & Line Charts were created for Gender, Age, and Race-based patient distribution.
- Heatmaps were used to analyze wait time trends by week and hour.
- KDE Plot was used for age distribution.

6. Key Findings

- Satisfaction Scores are missing for over 70% of patients, making analysis difficult.
- General Practice is the most visited department, followed by Orthopedics.
- Most ER visits happened during weekdays, with peak hours in the morning and late evening.
- Men had slightly higher satisfaction scores than women.
- Older patients had lower satisfaction scores, possibly due to longer wait times.

7. Recommendations

- Improve Satisfaction Data Collection: Address missing satisfaction scores by requiring patients to provide feedback.
- Reduce Wait Times in Key Departments: Focus on Physiotherapy, Cardiology, and Neurology to improve efficiency.
- Improve Senior Patient Experience: Since older patients report lower satisfaction, reducing wait times for them can help.
- Further Analysis on "NC" Gender Category: This group had significantly lower satisfaction (3.20), requiring further investigation.

Importing Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

Importing File From the Location

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	Date	Patient ID	Gender	Age	Satisfaction Score	First Name	Last Name	Race	Patic Adn F
0	20/03/2020 8:47	145- 39- 5406	М	69	10.0	Н	Glasspool	White	Fa
1	15/06/2020 11:29	316- 34- 3057	М	4	NaN	Х	Methuen	Native American/Alaska Native	Т
2	20/06/2020 9:13	897- 46- 3852	F	56	9.0	Р	Schubuser	African American	Т
3	04/02/2020 22:34	358- 31- 9711	F	24	8.0	U	Titcombe	Native American/Alaska Native	Т
4	04/09/2020 17:48	289- 26- 0537	М	5	NaN	Υ	Gionettitti	African American	Fa

Checking the shape of the data

```
In [3]: df.shape
```

Out[3]: (9216, 11)

Checking null values

```
In [4]: df.isnull().sum()
Out[4]: Date
                                  0
        Patient ID
                                  0
        Gender
                                  0
        Age
        Satisfaction Score
                             6699
        First Name
        Last Name
                                  0
        Race
        Patient Admin Flag
                                  0
        Wait Time
        Department Referral
                             5400
        dtype: int64
In [5]: Satisfaction_Score = 6699
        Department_Referral = 5400
        def per(x):
            return (x/len(df))*100
```

```
print("Satisfaction Score:- ", per(Satisfaction_Score))
print("Department Referral:- ", per(Department_Referral))
Satisfaction Score:- 72 68880208333334
```

Satisfaction Score:- 72.68880208333334

Department Referral: - 58.59375

Null values will not be deleted but using existing values analysis will be performed

Checking duplicates

```
In [6]: df.duplicated().sum()
Out[6]: np.int64(0)
```

Checking data types

```
df.dtypes
In [7]:
                               object
Out[7]: Date
        Patient ID
                               object
        Gender
                               object
                               int64
        Age
        Satisfaction Score float64
        First Name
                             object
        Last Name
                              object
                               object
        Race
        Patient Admin Flag
                                bool
                               int64
        Wait Time
        Department Referral
                              object
        dtype: object
```

Converting "Date" Column in datetime datatype

```
In [8]: df['Date'] = pd.to_datetime(df['Date'])
In [9]: df.info() #checking data type
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9216 entries, 0 to 9215
Data columns (total 11 columns):

	•	,	
#	Column	Non-Null Count	Dtype
0	Date	9216 non-null	datetime64[ns]
1	Patient ID	9216 non-null	object
2	Gender	9216 non-null	object
3	Age	9216 non-null	int64
4	Satisfaction Score	2517 non-null	float64
5	First Name	9216 non-null	object
6	Last Name	9216 non-null	object
7	Race	9216 non-null	object
8	Patient Admin Flag	9216 non-null	bool
9	Wait Time	9216 non-null	int64
10	Department Referral	3816 non-null	object

dtypes: bool(1), datetime64[ns](1), float64(1), int64(2), object(6)

memory usage: 729.1+ KB

Extracting "Time" from date

In [10]: df['Time'] = df['Date'].dt.time

In [11]: df.head() #checking data

Out[11]:

	Date	Patient ID	Gender	Age	Satisfaction Score	First Name	Last Name	Race	Patient Admin Flag
0	2020- 03-20 08:47:00	145- 39- 5406	М	69	10.0	Н	Glasspool	White	False
1	2020- 06-15 11:29:00	316- 34- 3057	М	4	NaN	Х	Methuen	Native American/Alaska Native	True
2	2020- 06-20 09:13:00	897- 46- 3852	F	56	9.0	Р	Schubuser	African American	True
3	2020- 02-04 22:34:00	358- 31- 9711	F	24	8.0	U	Titcombe	Native American/Alaska Native	True
4	2020- 09-04 17:48:00	289- 26- 0537	М	5	NaN	Υ	Gionettitti	African American	False

In [12]: df.dtypes #checking datatype

```
Out[12]: Date
                                datetime64[ns]
         Patient ID
                                        object
         Gender
                                        object
                                         int64
         Age
         Satisfaction Score
                                       float64
         First Name
                                        object
         Last Name
                                        object
         Race
                                        object
                                          bool
         Patient Admin Flag
                                         int64
         Wait Time
         Department Referral
                                        object
         Time
                                        object
         dtype: object
```

Converting "Time" format in time datatype

```
In [13]: df['Time'] = pd.to_datetime(df['Time'], format="%H:%M:%S")

In [14]: df['Hour'] = df['Time'].dt.hour #extracting hour from time colums
    df['Year'] = df['Date'].dt.year #extracting year from time columns
    df['Month'] = df['Date'].dt.month #extracting month from time columns
    df['Week_day_Number'] = df['Date'].dt.isocalendar().day #extracting Week day in num
    df['Days'] = df['Date'].dt.day #extracting Day from time columns
    df['Week_Name'] = df['Date'].dt.day_name() #extracting Week Name from time columns
    df['Month_Name'] = df['Date'].dt.month_name() #extracting month Name from time columns
```

In [15]: df.head() #checking data

Out[15]:

	Date	Patient ID	Gender	Age	Satisfaction Score	First Name	Last Name	Race	Patient Admin Flag
0	2020- 03-20 08:47:00	145- 39- 5406	М	69	10.0	Н	Glasspool	White	False
1	2020- 06-15 11:29:00	316- 34- 3057	М	4	NaN	Х	Methuen	Native American/Alaska Native	True
2	2020- 06-20 09:13:00	897- 46- 3852	F	56	9.0	Р	Schubuser	African American	True
3	2020- 02-04 22:34:00	358- 31- 9711	F	24	8.0	U	Titcombe	Native American/Alaska Native	True
4	2020- 09-04 17:48:00	289- 26- 0537	М	5	NaN	Υ	Gionettitti	African American	False

concatenating "First Name" & "Last Name" Columns As "Full Name"

In [16]: df['Full_Name'] = df['First Name']+'. '+df['Last Name']

In [17]: df.head() #checking data

Out[17]:

•		Date	Patient ID	Gender	Age	Satisfaction Score	First Name	Last Name	Race	Patient Admin Flag
	0	2020- 03-20 08:47:00	145- 39- 5406	М	69	10.0	Н	Glasspool	White	False
	1	2020- 06-15 11:29:00	316- 34- 3057	М	4	NaN	Х	Methuen	Native American/Alaska Native	True
	2	2020- 06-20 09:13:00	897- 46- 3852	F	56	9.0	Р	Schubuser	African American	True
	3	2020- 02-04 22:34:00	358- 31- 9711	F	24	8.0	U	Titcombe	Native American/Alaska Native	True
	4	2020- 09-04 17:48:00	289- 26- 0537	М	5	NaN	Υ	Gionettitti	African American	False

Droping Columns "First Name" & "Last Name"

In [18]: df.drop(columns=['First Name','Last Name'], inplace=True, axis=1)
 df.sample(5) #Checking data rendomly

Out[18]:

	Date	Patient ID	Gender	Age	Satisfaction Score	Race	Patient Admin Flag	Wait Time	Departi Ref
785	2020- 05-25 03:29:00	852- 96- 3176	М	56	NaN	Two or More Races	True	34	
2689	2019- 04-30 04:41:00	218- 44- 6635	М	36	NaN	Native American/Alaska Native	False	13	Ge Pra
6318	2019- 12-08 22:00:00	390- 43- 5551	F	14	7.0	White	False	41	Ge Pra
6630	2020- 05-29 10:35:00	883- 97- 1220	F	16	2.0	Two or More Races	True	24	
327	2019- 04-17 09:04:00	782- 34- 4167	F	36	9.0	White	True	50	Ge Pra

Finding Count or visiting of patient for each categories

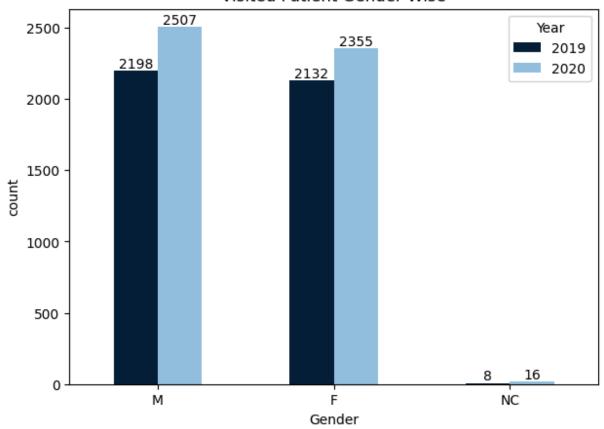
```
In [19]: for i in ['Gender', 'Race', 'Department Referral','Patient Admin Flag']:
    print(df[i].value_counts())
    print("-----")
```

```
Gender
M 4705
F
    4487
NC
     24
Name: count, dtype: int64
-----
White
                       2571
African American
                      1951
Two or More Races
                      1557
Asian
                      1060
Declined to Identify
                      1030
                       549
Pacific Islander
Native American/Alaska Native 498
Name: count, dtype: int64
-----
Department Referral
General Practice 1840
Orthopedics
Physiotherapy
            276
Cardiology
             248
              193
Neurology
Gastroenterology
             178
Renal
              86
Name: count, dtype: int64
-----
Patient Admin Flag
True 4612
False 4604
Name: count, dtype: int64
```

Performing EDA

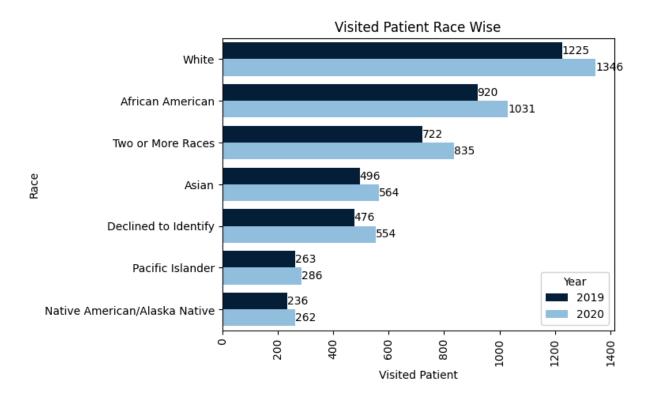
```
In [20]: plt.figure(figsize=(7,5)) #adust figure size
   plt.title('Visited Patient Gender Wise') #Title
   ax = sns.countplot(data=df, x = 'Gender', hue = 'Year', width=0.5, palette=['#001F
   ax.bar_label(ax.containers[0], fontsize=10)
   ax.bar_label(ax.containers[1], fontsize=10)
   plt.show()
```

Visited Patient Gender Wise



Out[21]:		Year	Race	Visited Patient
	0	2019	White	1225
	1	2019	African American	920
	2	2019	Two or More Races	722
	3	2019	Asian	496
	4	2019	Declined to Identify	476
	5	2019	Pacific Islander	263
	6	2019	Native American/Alaska Native	236
	7	2020	White	1346
	8	2020	African American	1031
	9	2020	Two or More Races	835
	10	2020	Asian	564
	11	2020	Declined to Identify	554
	12	2020	Pacific Islander	286
	13	2020	Native American/Alaska Native	262

```
In [22]: plt.title('Visited Patient Race Wise')
    ax = sns.barplot(data = r1, y = 'Race', x = 'Visited Patient', hue = 'Year', palett
    ax.bar_label(ax.containers[0], fontsize=10)
    ax.bar_label(ax.containers[1], fontsize=10)
    plt.xticks(rotation = 90)
    plt.show()
```



In [23]: df.groupby(df['Year'])['Patient Admin Flag'].value_counts().reset_index()

Out[23]: Year Patient Admin Flag count

0	2019	False	2181
1	2019	True	2157
2	2020	True	2455
3	2020	False	2423

```
In [24]: plt.title('Total Visited Patient Admin Flag Wise')

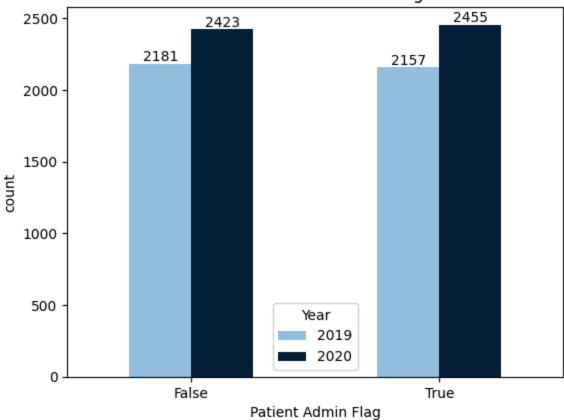
# Define custom golden & dark blue colors
custom_palette = ['#85C1E9', '#001F3F']

# Create bar chart
ax = sns.countplot(data=df, x='Patient Admin Flag', hue='Year', width=0.5, palette=

# Add value labels to bars
for container in ax.containers:
    ax.bar_label(container, fontsize=10)

plt.show()
```

Total Visited Patient Admin Flag Wise



```
In [25]: data_2019 = df[df['Year']==2019].groupby(['Gender'])['Department Referral'].value_c
    data_2019.rename({'count':'Patient_visit'}, axis= 1 , inplace = True)
    data_2020 = df[df['Year']==2020].groupby(['Gender'])['Department Referral'].value_c
    data_2020.rename({'count':'Patient_visit'}, axis= 1 , inplace = True)
    data_2020
    #data_2019, data_2020
```

	Gender	Department Referral	Patient_visit
0	F	General Practice	473
1	F	Orthopedics	238
2	F	Physiotherapy	71
3	F	Cardiology	59
4	F	Neurology	57
5	F	Gastroenterology	42
6	F	Renal	20
7	М	General Practice	488
8	М	Orthopedics	293
9	М	Physiotherapy	71
10	М	Cardiology	64
11	М	Neurology	47
12	М	Gastroenterology	42
13	М	Renal	20
14	NC	General Practice	5
15	NC	Gastroenterology	1
16	NC	Orthopedics	1

Out[25]:

```
In [26]: plt.figure(figsize=(15,6))
         # Subplot for 2019
         plt.subplot(1,2,1)
         plt.title('Department Referral Wise Visit In 2019')
         ax1 = sns.lineplot(data=data_2019, x='Department Referral', y='Patient_visit', hue=
         plt.xticks(rotation=90)
         # Add value labels
         for line in ax1.lines:
             for x, y in zip(line.get_xdata(), line.get_ydata()):
                 plt.text(x, y, f"{int(y)}", ha='right', va='bottom', fontsize=8)
         # Subplot for 2020
         plt.subplot(1,2,2)
         plt.title('Department Referral Wise Visit In 2020')
         ax2 = sns.lineplot(data=data_2020, x='Department Referral', y='Patient_visit', hue=
         plt.xticks(rotation=90)
         # Add value labels
         for line in ax2.lines:
             for x, y in zip(line.get_xdata(), line.get_ydata()):
```

```
plt.text(x, y, f"{int(y)}", ha='right', va='bottom', fontsize=8)

plt.show()

Department Referral Wise Visit In 2019

Department Referral Wise Visit In 2020

Gender

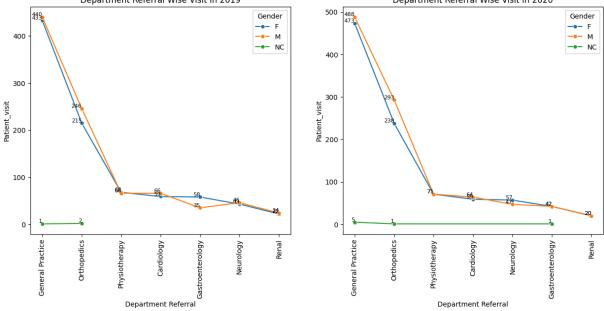
--- F

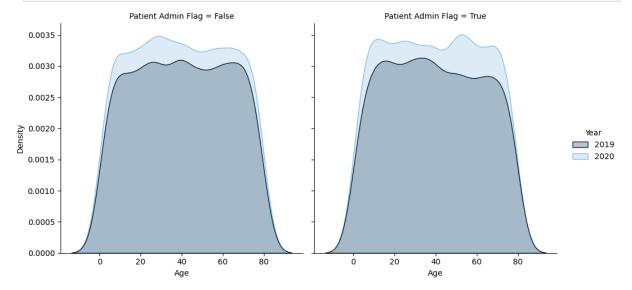
--- M

Gender

--- F

--- M
```





In [28]: round(pd.pivot_table(data = df, index='Department Referral', columns='Gender', valu

Out[28]:	Gender	F	M	NC
	Department Referral			
	Cardiology	35.1	35.6	NaN
	Gastroenterology	34.5	37.9	10.0
	General Practice	35.1	34.7	39.2
	Neurology	36.7	36.9	NaN
	Orthopedics	34.5	35.4	30.0
	Physiotherapy	36.0	37.2	NaN
	Renal	35.5	33.9	NaN

```
In [29]: from matplotlib.colors import LinearSegmentedColormap

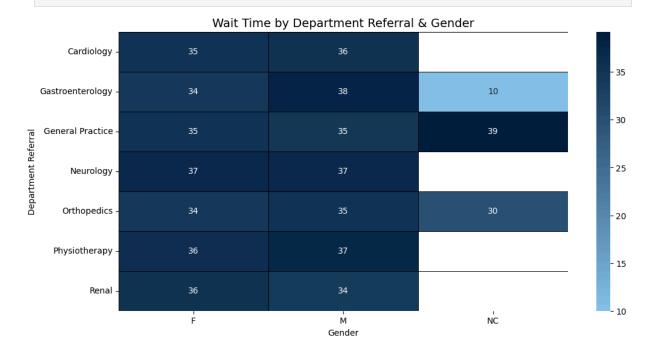
# Create a custom colormap from dark blue to gold
golden_blue_cmap = LinearSegmentedColormap.from_list("golden_blue", ["#85C1E9", "#0

# Create pivot table and round values
pivot_table = round(pd.pivot_table(data=df, index='Department Referral', columns='G

# Plot heatmap
plt.figure(figsize=(12,6))
sns.heatmap(pivot_table, annot=True, cmap=golden_blue_cmap, linewidths=0.5, linecol

# Title
plt.title('Wait Time by Department Referral & Gender', fontsize=14)

plt.show()
```



```
In [30]: from matplotlib.colors import LinearSegmentedColormap

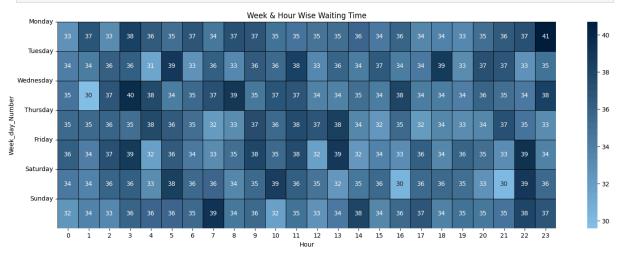
plt.figure(figsize=(18,6))
plt.title('Week & Hour Wise Waiting Time')

# Create a golden & dark blue colormap
golden_blue_cmap = LinearSegmentedColormap.from_list("golden_blue", ["#85C1E9", "#0

# Pivot table
pivot_r1 = round(pd.pivot_table(data=df, index='Week_day_Number', columns='Hour', v

# Plot heatmap
sns.heatmap(pivot_r1, annot=True, cmap=golden_blue_cmap, linewidths=0.5, linecolor=

# Customize y-axis labels (weekdays)
week_labels = ['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', plt.yticks(ticks=range(len(week_labels)), labels=week_labels, rotation=0, ha='right
plt.show()
```



```
In [31]: df['Age'].value_counts()
Out[31]: Age
          39
                 145
          49
                 142
          27
                 142
          18
                 139
          56
                 139
                . . .
          20
                 101
          1
                 101
          75
                 100
          44
                 100
                 100
          Name: count, Length: 79, dtype: int64
In [32]: def person(x):
```

if x <18:

return 'Children'
elif x > 18 and x<=45:</pre>

return 'Adults'
else:
return 'Seniors'

In [33]: df['Age_Group'] = df['Age'].apply(person)

In [34]: df.sample(4)

Out[34]:

	Date	Patient ID	Gender	Age	Satisfaction Score	Race	Patient Admin Flag	Wait Time	Departi Ref
778	2019- 1 10-15 21:26:00	164- 06- 4233	F	59	NaN	Pacific Islander	True	43	Orthop
295	2019- 1 12-27 22:28:00	593- 38- 7471	F	62	NaN	African American	True	32	Orthop
750	2019- 6 05-04 15:38:00	869- 06- 1488	М	8	NaN	Two or More Races	True	27	
215	2019- 3 07-28 18:20:00	488- 25- 0630	F	73	NaN	Native American/Alaska Native	False	18	

In [35]: s1 = round(df.groupby('Age_Group')['Satisfaction Score'].mean(),2)
s1

Out[35]: Age_Group

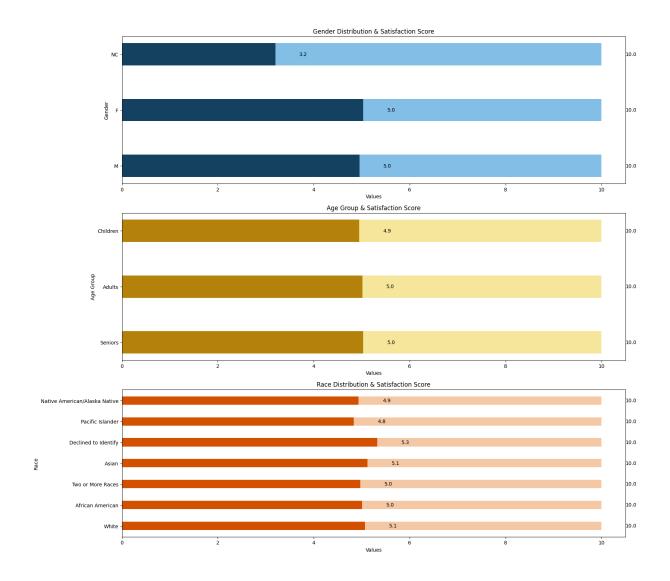
Adults 5.03 Children 5.01 Seniors 4.95

Name: Satisfaction Score, dtype: float64

In [36]: df[['Age', 'Age_Group']]

```
Out[36]:
                Age Age_Group
             0
                 69
                         Seniors
                  4
                        Children
             2
                 56
                         Seniors
                          Adults
                 24
             4
                  5
                        Children
          9211
                 60
                         Seniors
                 63
          9212
                         Seniors
          9213
                          Adults
                 27
          9214
                 53
                         Seniors
          9215
                 30
                          Adults
         9216 rows × 2 columns
In [37]: s2 = round(df.groupby('Gender')['Satisfaction Score'].mean(),2)
          s2
Out[37]: Gender
                4.96
          Μ
                5.03
          NC
                3.20
          Name: Satisfaction Score, dtype: float64
In [38]: s3 = round(df.groupby('Race')['Satisfaction Score'].mean(),2)
          s3
Out[38]: Race
                                            5.07
          African American
          Asian
                                            5.01
          Declined to Identify
                                            4.97
          Native American/Alaska Native
                                            5.12
          Pacific Islander
                                            5.33
          Two or More Races
                                            4.83
          White
                                            4.94
          Name: Satisfaction Score, dtype: float64
In [39]: plt.figure(figsize=(18, 18))
          # ---- Gender Distribution ----
          plt.subplot(3,1,1)
          gender_counts = df['Gender'].value_counts()
          satisfaction_scores = df.groupby('Gender')['Satisfaction Score'].mean().values
          ax1 = plt.barh(y=gender_counts.index, width=[10,10,10], color='#85C1E9', height=0.4
          ax2 = plt.barh(y=gender_counts.index, width=satisfaction_scores, color='#154360', h
```

```
for bars in [ax1, ax2]:
   for bar in bars:
        plt.text(bar.get_width() + 0.5, bar.get_y() + bar.get_height()/2, f"{bar.ge
                 va='center', fontsize=10, color='black')
plt.xlabel("Values")
plt.ylabel("Gender")
plt.title("Gender Distribution & Satisfaction Score")
# ---- Age Group Distribution ----
plt.subplot(3,1,2)
age_counts = df['Age_Group'].value_counts()
satisfaction_scores = df.groupby('Age_Group')['Satisfaction Score'].mean().values
ax1 = plt.barh(y=age_counts.index, width=[10,10,10], color='#F9E79F', height=0.4, l
ax2 = plt.barh(y=age_counts.index, width=satisfaction_scores, color='#B8860B', heig
for bars in [ax1, ax2]:
   for bar in bars:
        plt.text(bar.get_width() + 0.5, bar.get_y() + bar.get_height()/2, f"{bar.ge
                 va='center', fontsize=10, color='black')
plt.xlabel("Values")
plt.ylabel("Age Group")
plt.title("Age Group & Satisfaction Score")
# ---- Race Distribution ----
plt.subplot(3,1,3)
race_counts = df['Race'].value_counts()
satisfaction_scores = df.groupby('Race')['Satisfaction Score'].mean().values
ax1 = plt.barh(y=race_counts.index, width=[10,10,10,10,10,10], color='#F5CBA7',
ax2 = plt.barh(y=race_counts.index, width=satisfaction_scores, color='#D35400', hei
for bars in [ax1, ax2]:
   for bar in bars:
        plt.text(bar.get_width() + 0.5, bar.get_y() + bar.get_height()/2, f"{bar.ge
                 va='center', fontsize=10, color='black')
plt.xlabel("Values")
plt.ylabel("Race")
plt.title("Race Distribution & Satisfaction Score")
plt.show()
```



Summary of the Hospital Emergency Room Visit Analysis

This report analyzes 9,216 emergency room visits from a hospital dataset. The data includes patient demographics, wait times, department referrals, and satisfaction scores. Below are the key insights:

1. Data Overview

Total Records: 9,216 Columns: 11 (Date, Patient ID, Gender, Age, Satisfaction Score, Race, Wait Time, Department Referral, etc.) Missing Values: Satisfaction Score: 72.7% missing (6,699 entries) Department Referral: 58.6% missing (5,400 entries)

2. Patient Demographics

Gender Distribution:

Male: 4,705 Female: 4,487 Non-Classified (NC): 24

Race Breakdown:

White: 2,571 (largest group) African American: 1,951 Two or More Races: 1,557 Asian: 1,060 Other categories include: Pacific Islander, Native American, Declined to Identify

3. Emergency Room Visit Trends

Most Common Department Referrals:

General Practice: 1,840 visits Orthopedics: 995 visits Physiotherapy: 276 visits Others: Cardiology, Neurology, Gastroenterology, and Renal Yearly Breakdown:

2019: White: 1,225 visits African American: 920 visits 2020: White: 1,346 visits African American: 1,031 visits Admin Flag (Patient Admitted or Not):

True: 4,612 False: 4,604

4. Wait Time & Satisfaction Analysis

Wait Time Distribution (Grouped by Department & Gender)
On average, wait times were higher in Physiotherapy & Cardiology Males and Females had similar wait times across departments Satisfaction Score by Category:

By Age Group: Adults: 5.03 Children: 5.01 Seniors: 4.95 By Gender: Male: 5.03 Female: 4.96 Non-Classified (NC): 3.20 By Race: Highest Satisfaction: Pacific Islander (5.33), Native American (5.12) Lowest Satisfaction: Two or More Races (4.83), White (4.94)

5. Visualizations & Data Processing

Bar Charts & Line Charts were created for Gender, Age, and Race-based patient distribution. Heatmaps were used to analyze wait time trends by week and hour. KDE Plot was used for age distribution.

6. Key Findings

Satisfaction Scores are missing for over 70% of patients, making analysis difficult. General Practice is the most visited department, followed by Orthopedics. Most ER visits happened

during weekdays, with peak hours in the morning and late evening. Men had slightly higher satisfaction scores than women. Older patients had lower satisfaction scores, possibly due to longer wait times.

7. Recommendations

Improve Satisfaction Data Collection: Address missing satisfaction scores by requiring patients to provide feedback. Reduce Wait Times in Key Departments: Focus on Physiotherapy, Cardiology, and Neurology to improve efficiency. Improve Senior Patient Experience: Since older patients report lower satisfaction, reducing wait times for them can help. Further Analysis on "NC" Gender Category: This group had significantly lower satisfaction (3.20), requiring further investigation.

In []: