

A Comprehensive Data Analysis on Medical Expenses



# Medical Cost Analysis Presentation

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# Basic Statistics And Data Understanding

Data Columns	AVG	Median	STDEV
Charges	13279.12	9386.16	12110.36
BMI	30.66	30.40	6.10
Age	39.22	39.00	14.04

## ➤ Overview of Key Metrics

To gain an initial understanding of the dataset, we analyzed key statistical measures for **BMI**, **Age** and **Medical Charges**. These metrics offer insights into the **distribution** and **variability** of health-related factors across individuals.

## ➤ Average Values and Trends

The **average BMI** is **30.66**, indicating a trend toward **overweight or obesity** among individuals in the dataset. The **average age** is **39.21**, showing a mix of **younger and older** individuals. Additionally, the **average medical charges** amount to **\$13,270.42**, reflecting **significant healthcare expenses**.

## ➤ Central Tendency and Dispersion

The **median values** for **BMI (30.4)**, **age (39)**, and **charges (\$9,382.03)** provide insights into the **center of the data distribution**. The **standard deviation** for **BMI (6.10)**, **age (14.05)**, and **medical charges (\$12,110.01)** indicate **considerable variation**, particularly in medical costs where some individuals incur **substantially higher expenses**.

## ➤ Key Observations

The **high standard deviation in medical charges** suggests **significant disparities in healthcare costs**, likely influenced by factors such as **chronic conditions, insurance coverage, or hospitalization needs**. The **relatively close values of average and median age** indicate a **balanced distribution** across age groups. **BMI levels** further emphasize the importance of addressing **weight-related health risks** in the population.

# Gender-Based Analysis

## ➤ Objective

To compare the **average medical charges** and **BMI** between **male** and **female** beneficiaries.

## ➤ Method

Used AVERAGEIF and Pivot Table functions in Excel to calculate the **average BMI** and **average medical charges** for each gender.

## ➤ Findings

- **Average BMI (Male): 30.94**
- **Average BMI (Female): 30.38**
- **Average Charges (Male): \$13,975.00**
- **Average Charges (Female): \$12,569.58**

Row Labels	Average of BMI	Average of charges
female	30.38	12569.58
male	30.94	13975.00
Grand Total	30.66	13279.12

## ➤ Insights

- Males have both **higher average BMI** and **higher average medical charges** compared to females.
- The difference suggests a **potential correlation** between gender, BMI, and healthcare costs.

# Impact of Smoking on Medical Costs

## ➤ Objective

To evaluate how **smoking status** affects **average medical charges** among beneficiaries.

## ➤ Method

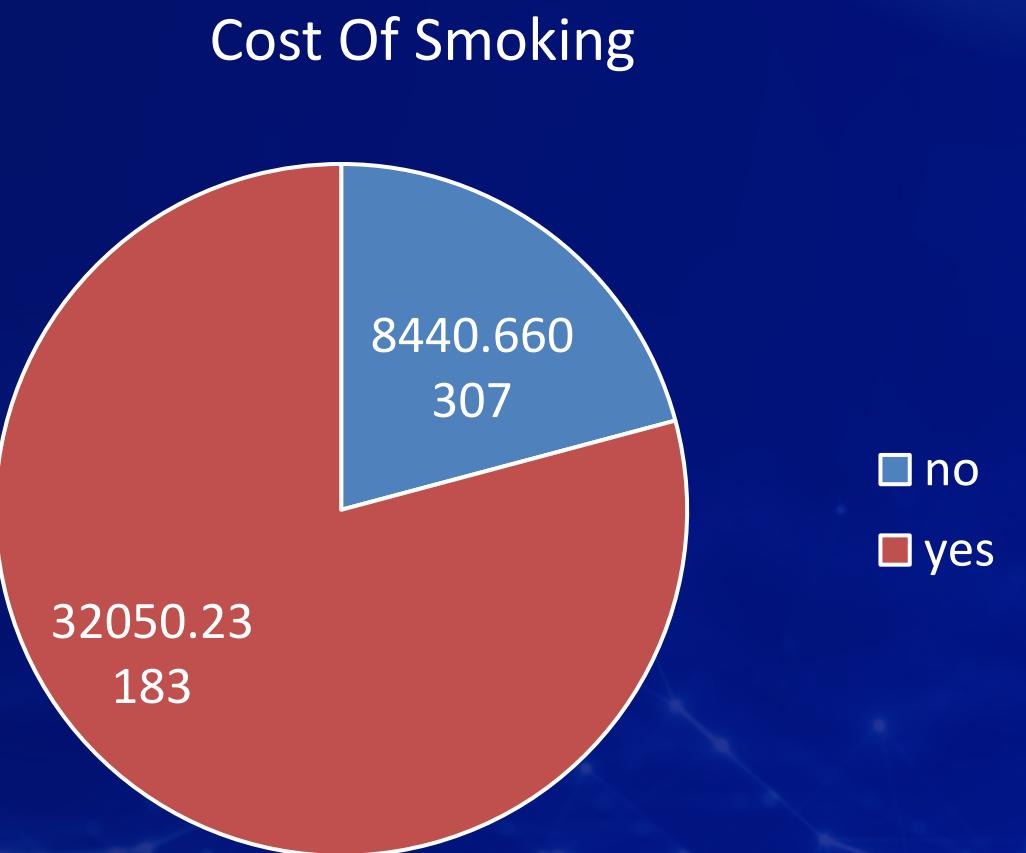
Used AVERAGEIFS and Pivot Table in Excel to calculate **average charges** separately for **smokers** and **non-smokers**.

## ➤ Findings

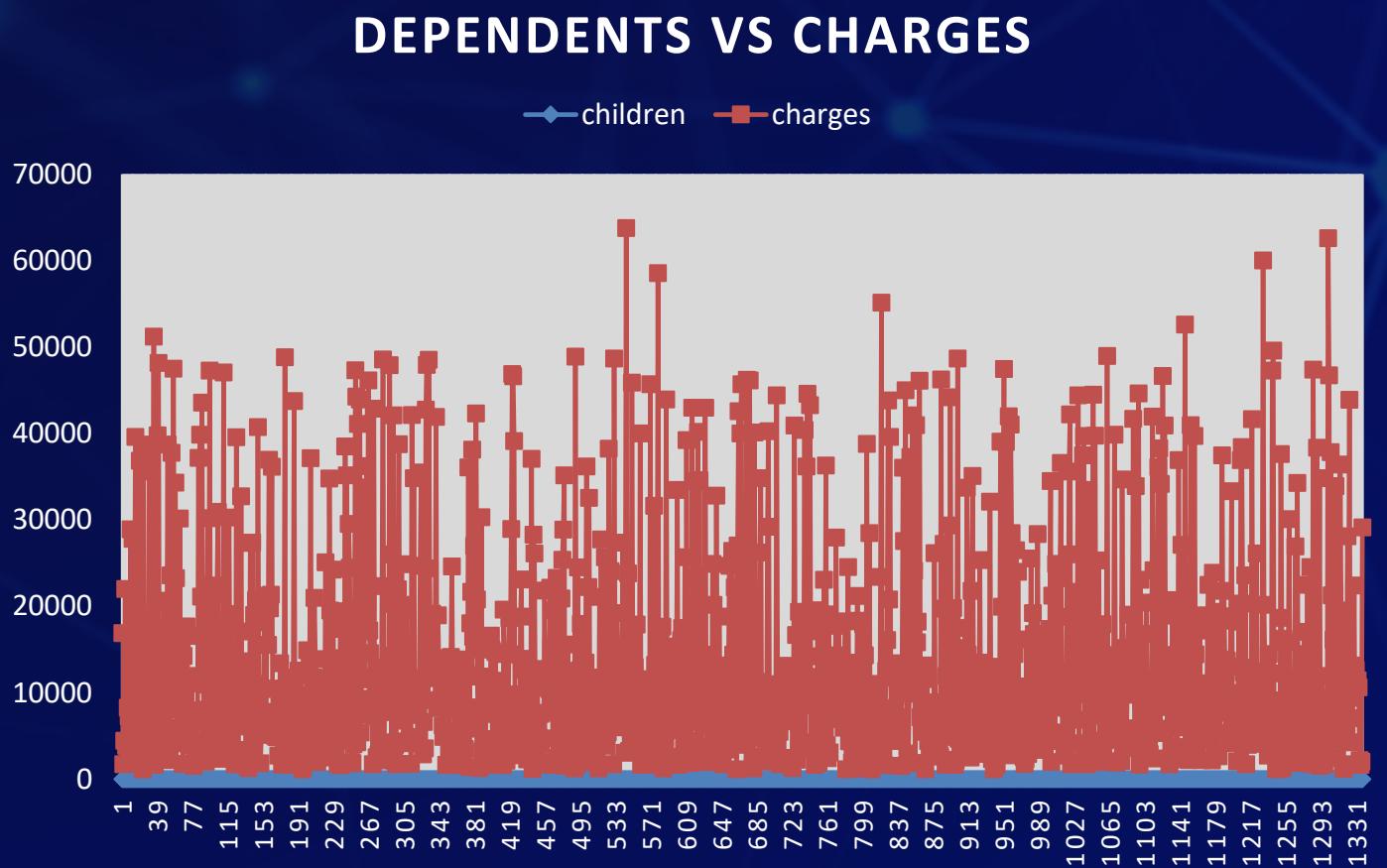
- **Average Charges (Smokers): \$32,050.23**
- **Average Charges (Non-Smokers): \$8,440.66**

## ➤ Insights

- **Smokers incur nearly 4 times higher** average medical costs compared to non-smokers.
- This significant cost difference highlights the **financial burden of smoking-related health issues**, making smoking a **major risk factor** in healthcare expenses.



# Dependents Analysis



## ➤ Objective

To evaluate whether the **number of children or dependents** influences **medical charges** among beneficiaries.

## ➤ Method

Used Excel's CORREL function to measure the statistical relationship between **children** and **charges**. Created a **line chart** to visualize this trend.

## ➤ Findings

➤ **Correlation coefficient** between number of children and medical charges is **0.067**, indicating a **very weak positive relationship**.

➤ The chart shows **no clear upward or downward trend**, with medical charges fluctuating widely across all dependent levels.

## ➤ Insights

➤ There is **no strong evidence** that having more children significantly increases medical costs.

➤ While a slight positive trend exists, it is **statistically insignificant**, suggesting that other factors (like smoking status, age, or chronic illness) may have a greater influence on charges.

# Geographical Insights

## ➤ Objective

To identify which **U.S. region** has the **highest average medical charges** and **BMI** among beneficiaries.

## ➤ Method

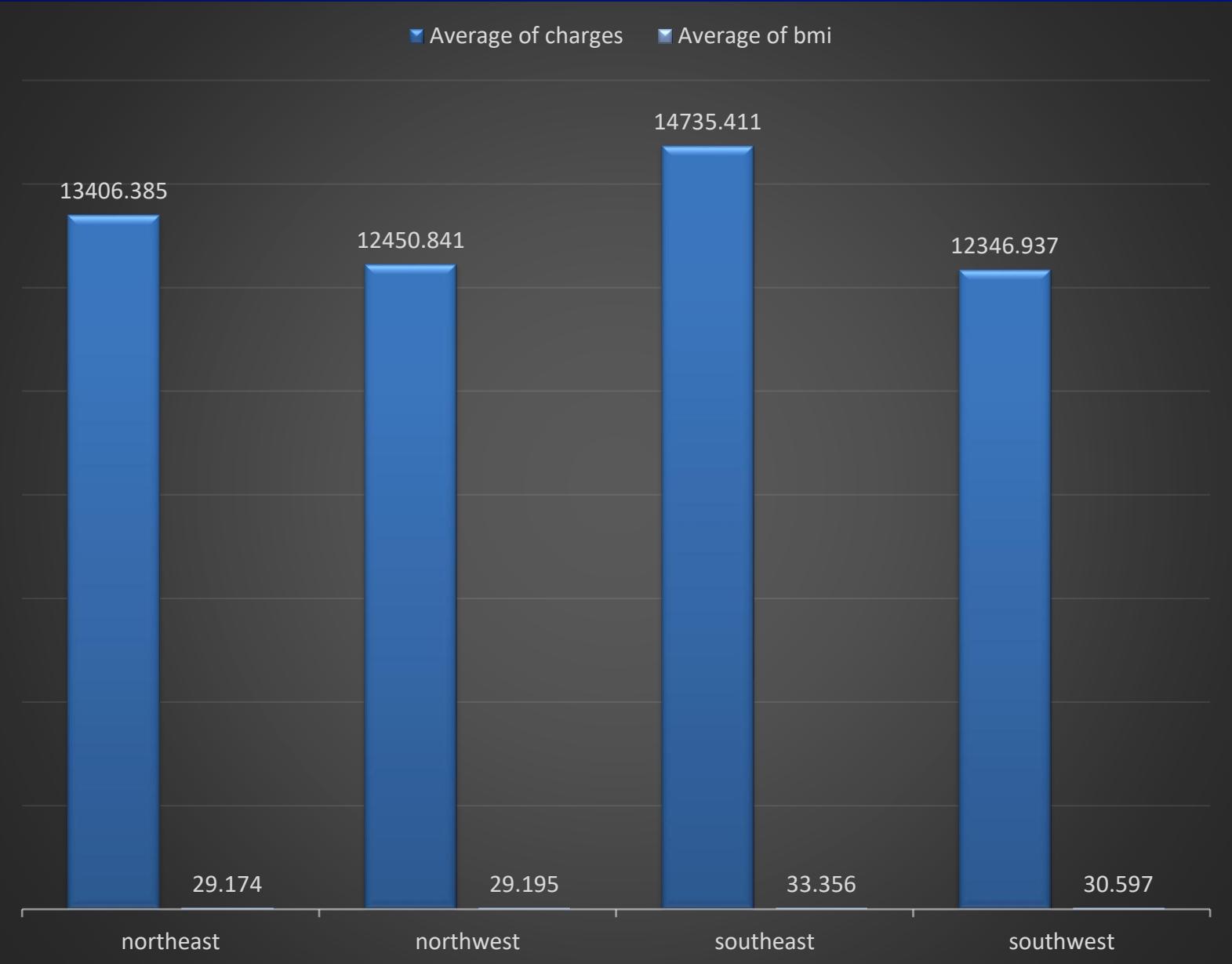
Used Excel Pivot Table and AVERAGEIFS to calculate **average charges** and **BMI** by **region**. A **bar chart** was created for visual comparison.

## ➤ Findings

- **Southeast** region has the **highest average medical charges** at **\$14,735.41**
- It also records the **highest average BMI** of **33.36**
- **Southwest** has the **lowest average charges** at **\$12,346.94**

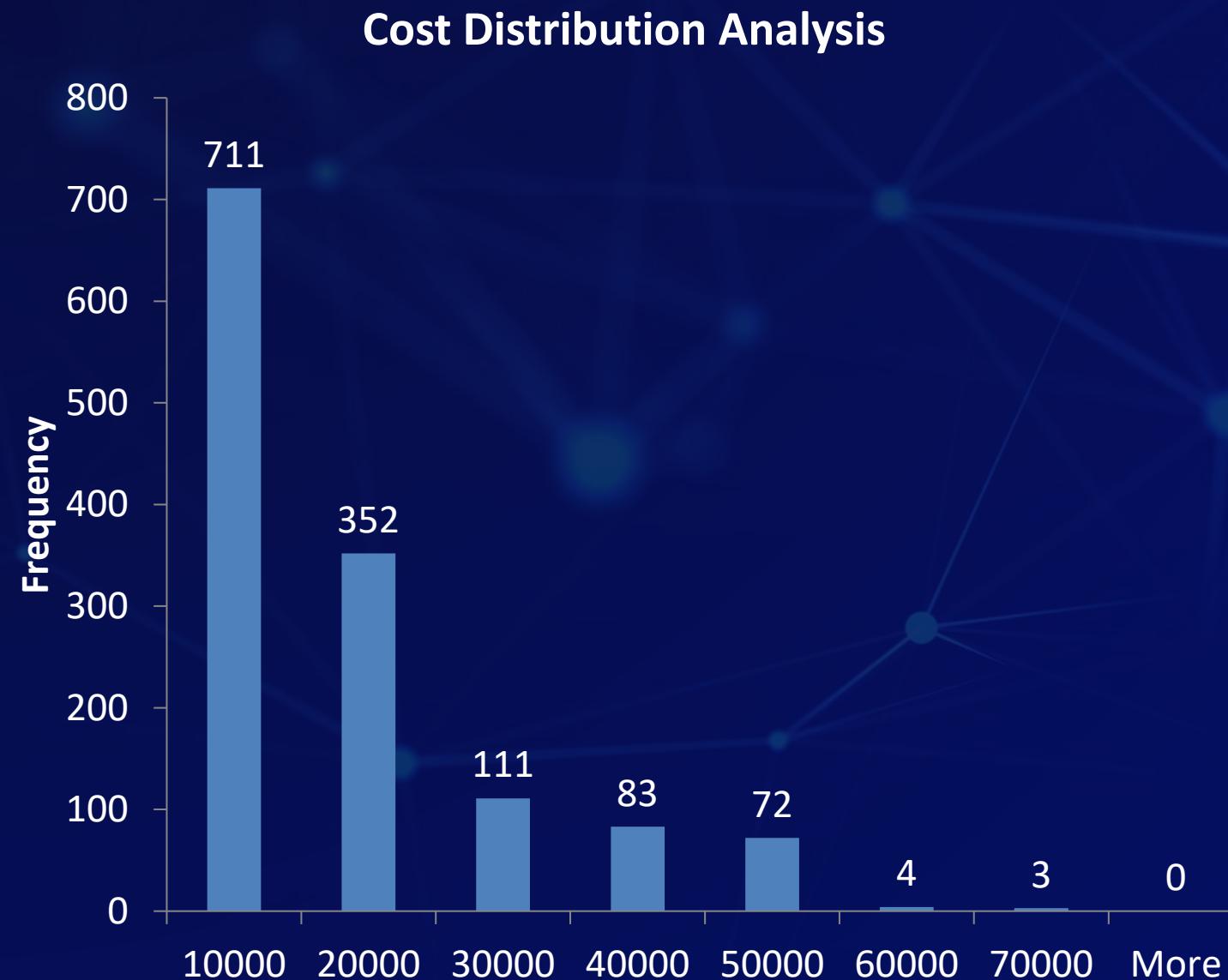
## ➤ Insights

- The **Southeast** region stands out with both **elevated BMI** and **higher healthcare costs**, indicating a **potential regional health risk pattern**.
- This suggests that **geographic location** may influence both **lifestyle** and **medical expenditures**, which could be useful for **targeted public health interventions**.





# Cost Distribution Analysis



## ➤ Objective

To understand how **medical charges** are distributed across beneficiaries and identify the **most common cost bracket**.

## ➤ Method

Used Excel's **Histogram** tool via the **Data Analysis Toolpak** along with the FREQUENCY function to group charges into defined **bin ranges**.

## ➤ Findings

➤ The **most common cost bracket** is **\$0–10,000**, with **711 beneficiaries** falling in this range.

➤ Frequency decreases significantly as charges increase:

➤ **\$10,001–20,000**: 352

➤ **\$20,001–30,000**: 111

➤ **Above \$30,000**: progressively fewer cases, only **3** above **\$70,000**

## ➤ Insights

➤ The data is **right-skewed**, with most beneficiaries incurring **lower medical expenses**.

➤ A **small number** of high-cost individuals create a **long tail** in the distribution, likely due to **critical illnesses or hospitalization**.

# BMI Classification

## ➤ Objective

To categorize beneficiaries based on their **BMI values** and evaluate the **distribution of weight categories** within the dataset.

## ➤ Method

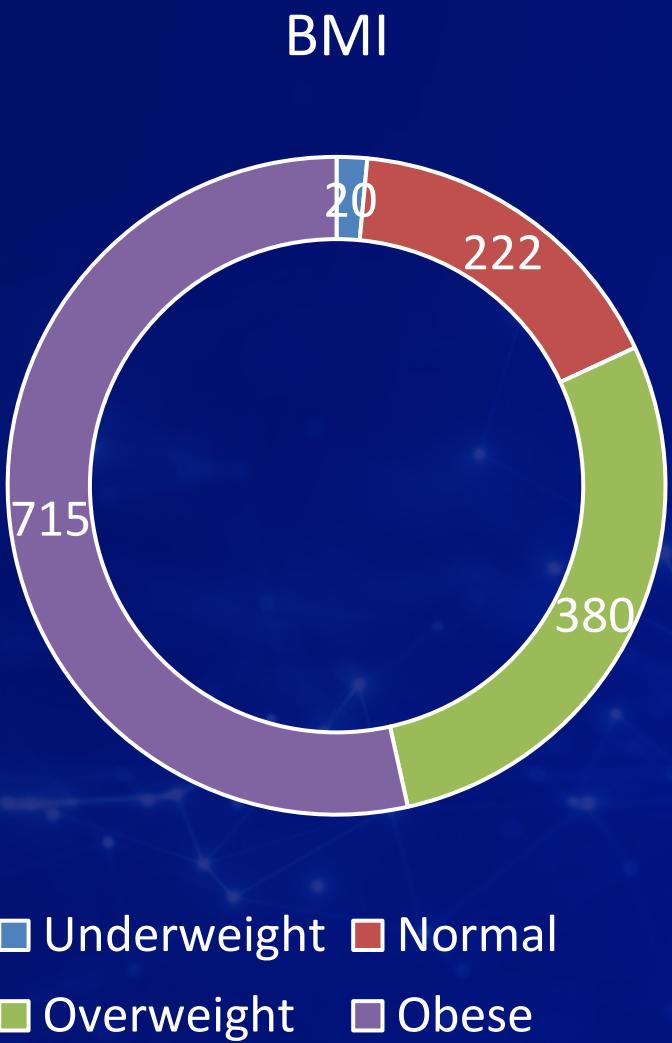
Applied Excel's COUNTIFS and VLOOKUP with **range-based lookup** to classify BMI into four categories: **Underweight**, **Normal**, **Overweight**, and **Obese**.

## ➤ Findings

- **Underweight (<18.5): 20 beneficiaries**
- **Normal (18.5–24.9): 225 beneficiaries**
- **Overweight (25–29.9): 386 beneficiaries**
- **Obese (30+): 706 beneficiaries**

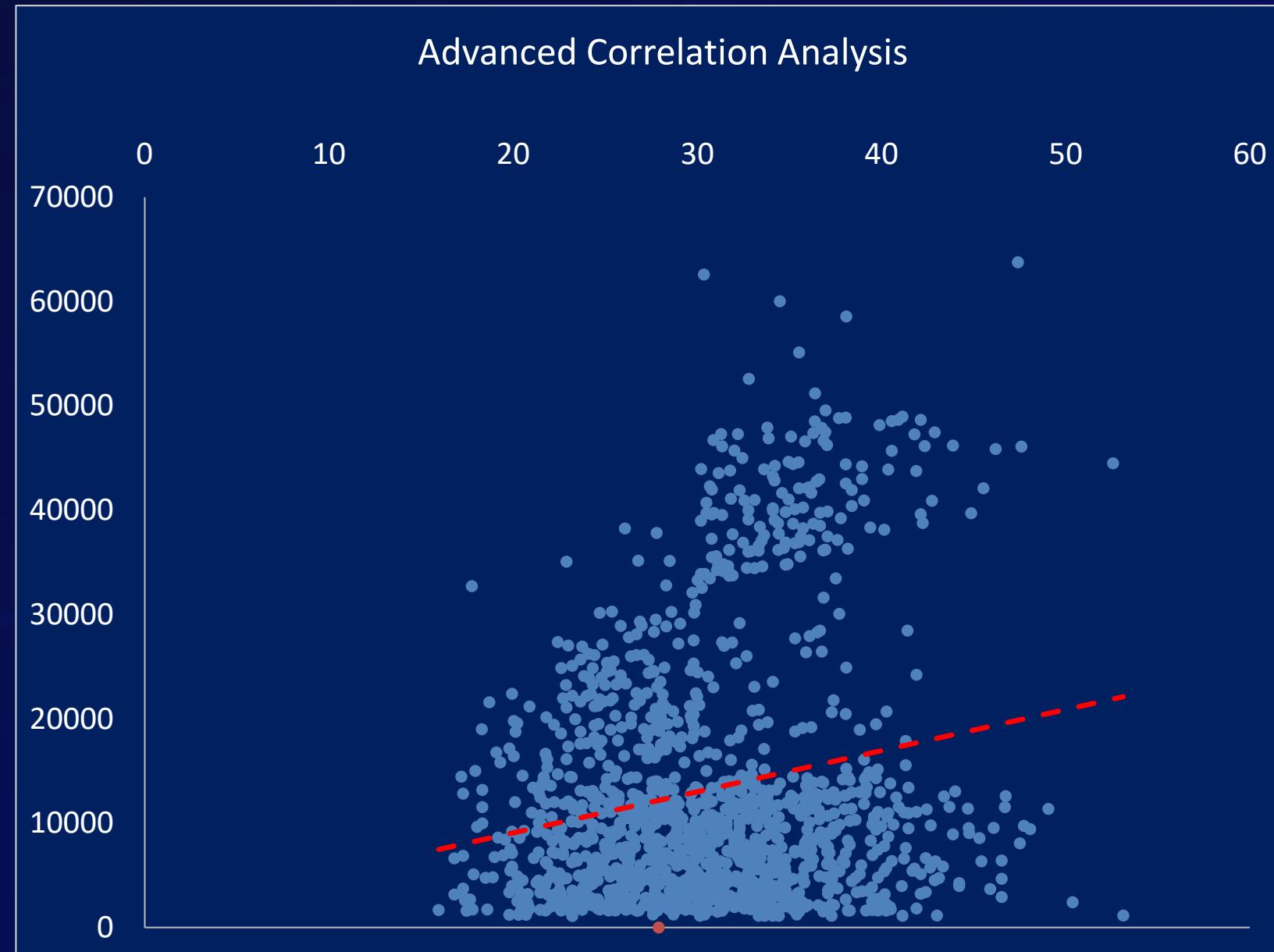
## ➤ Insights

- A **majority (706 individuals)** fall into the **Obese** category, followed by those who are **Overweight**.
- Only a **small portion** of the population is classified as **Underweight** or **Normal**, indicating a **potential public health concern** related to **high BMI levels**.
- These insights may support initiatives focused on **obesity prevention and management**.



# Advanced Correlation Analysis

Advanced Correlation Analysis



➤ **Objective**

To assess the **relationship between BMI and medical charges** using a **scatter plot** and **trendline**.

➤ **Method**

Created a **scatter plot** in Excel with **BMI on the x-axis** and **medical charges on the y-axis**. Added a **linear trendline** and calculated correlation using the CORREL function.

➤ **Findings**

➤ **Correlation coefficient** between **BMI and medical charges** is **0.198**, indicating a **weak positive relationship**.

➤ The **trendline** shows a **slight upward slope**, suggesting that higher BMI is **generally associated** with increased medical costs.

➤ **Insights**

➤ While the relationship is **not strong**, there is some evidence that **higher BMI levels** may contribute to **elevated medical expenses**.

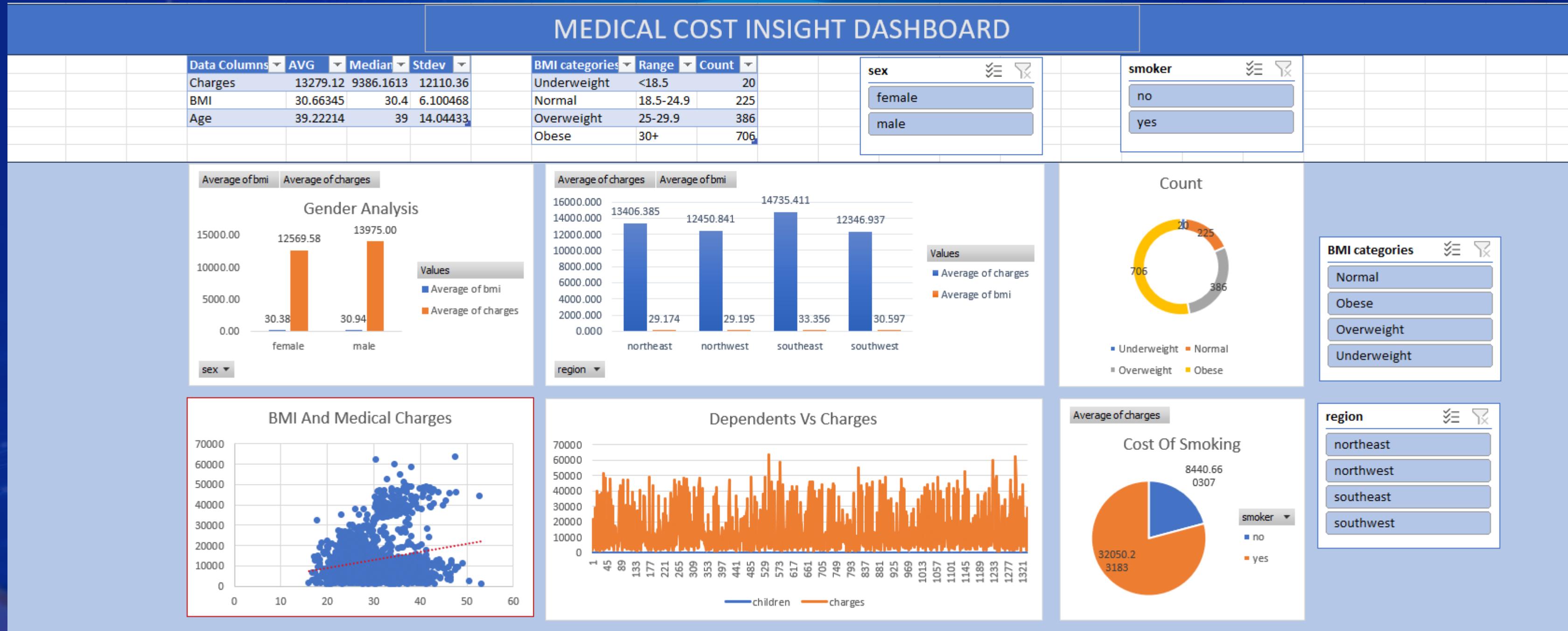
➤ This supports the need for **preventive healthcare** strategies targeting **weight-related health risks**.

# Smoking and Region Interaction

Sum of charges	Column Labels
Row Labels	yes
northeast	1988126.944
northwest	1751136.185
southeast	3170894.711
southwest	1871605.683
Grand Total	8781763.522

- **Objective**  
To explore how **smoking status and geographical region** together impact **medical charges**, and determine which region incurs the **highest smoker-related costs**.
- **Method**  
Created a **Pivot Table** in Excel with **region** as row labels and **smoking status** as a filter. Used AVERAGEIFS and **slicers** to isolate charges for **smokers only**.
- **Findings**
- **Southeast region shows the highest total medical charges** for smokers at **\$3,170,894.71**
- Other regions rank as follows:
  - **Northeast**: \$1,988,126.94
  - **Southwest**: \$1,871,605.68
  - **Northwest**: \$1,751,136.19
- **Insights**
- Smokers in the **Southeast** region account for a **significantly larger share** of healthcare expenses.
- This may reflect **regional health behaviors, access to care, or prevalence of smoking-related conditions**.
- Insights like these can help target **regional anti-smoking campaigns or healthcare resource planning**.

# Dashboard for Stakeholders





# Thank you

Data Speaks. Insights Lead.