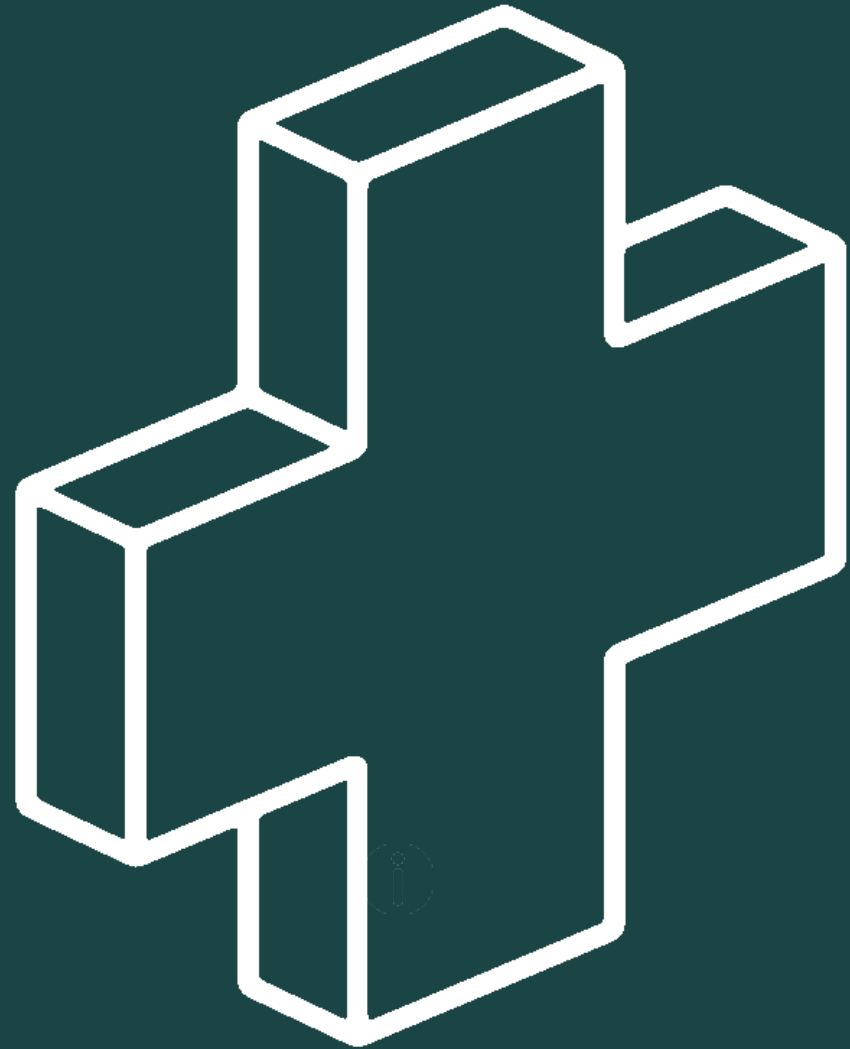




BUAN 5510 01- Capstone Project in  
Business Analytics



# PATIENT CHARACTERISTICS TO PREDICT THE TYPE OF HEALTHCARE SERVICE

Optimizing Healthcare Services through Predictive Analytics

**PRESENTED BY**

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# AGENDA

|   |                                |    |                              |
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| 1 | Introduction                   |    |                              |
| 2 | Problem Statement & Objectives |    |                              |
| 3 | Data Overview                  |    |                              |
| 4 | Key Findings & Insights        | 7  | Model Performance Evaluation |
| 5 | Data Preprocessing             | 8  | Predictive Insights          |
| 6 | Methodology                    | 9  | Use Cases and Stakeholders   |
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# INTRODUCTION



**Overview: Analyzing Patient Attributes for Healthcare Optimization.**

**Background Context: Increased diversity in patient needs, along with resource limitations.**

**Relevance: Enhancing Healthcare Efficiency through Predictive Insights.**



## PROBLEM STATEMENT

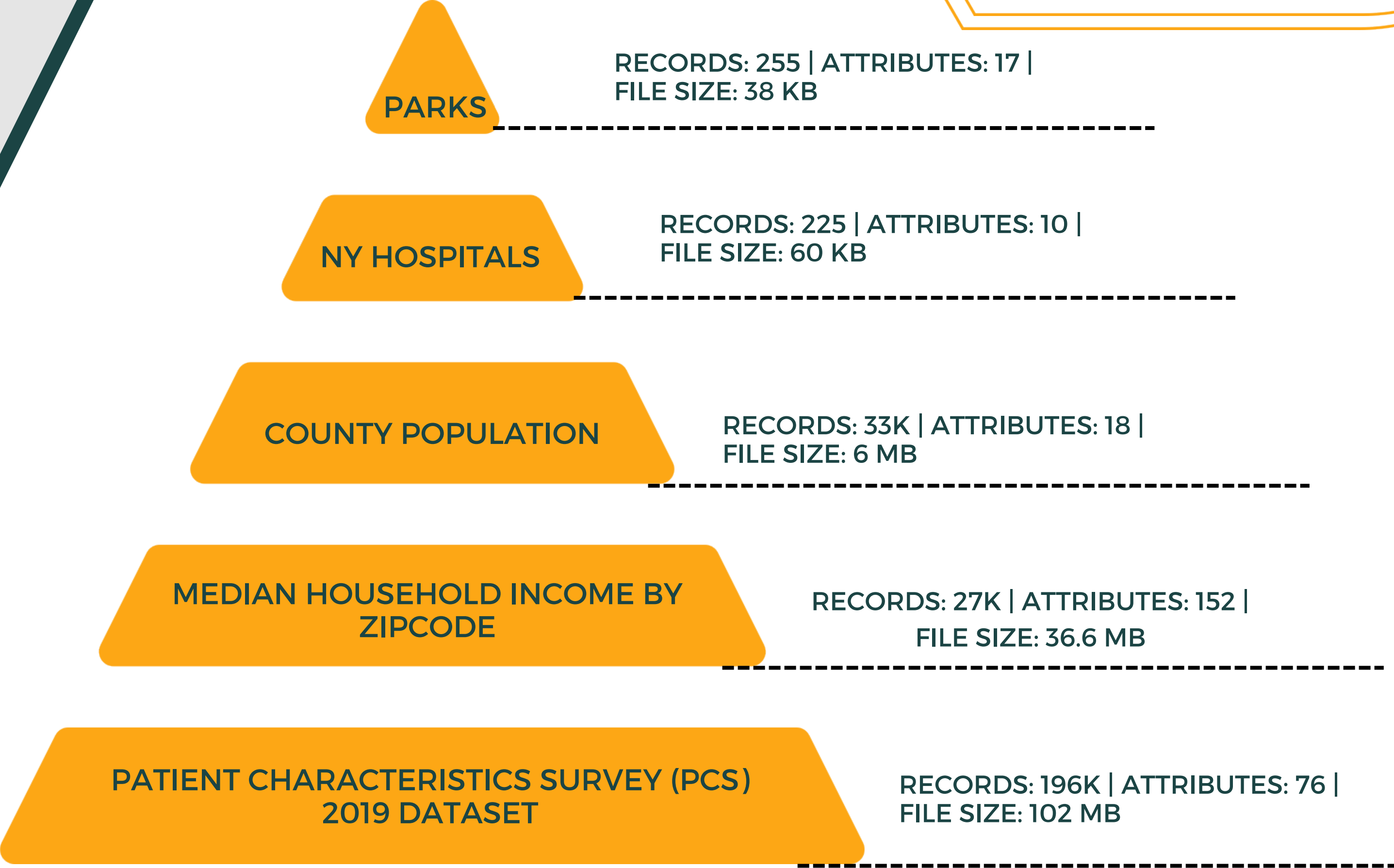
To predict the type of healthcare service (inpatient, outpatient, residential, emergency) patients are likely to use based on their characteristics and medical history

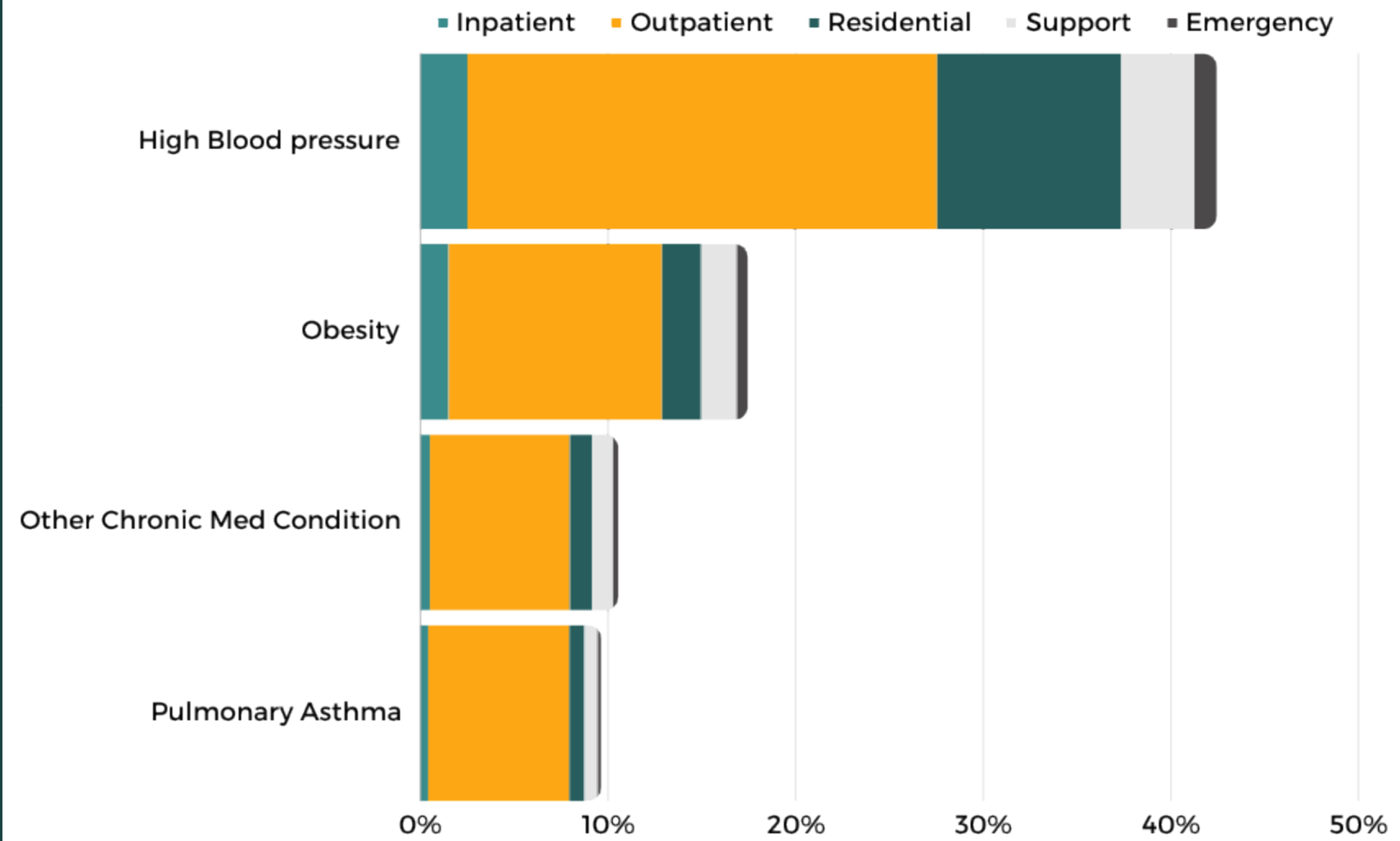
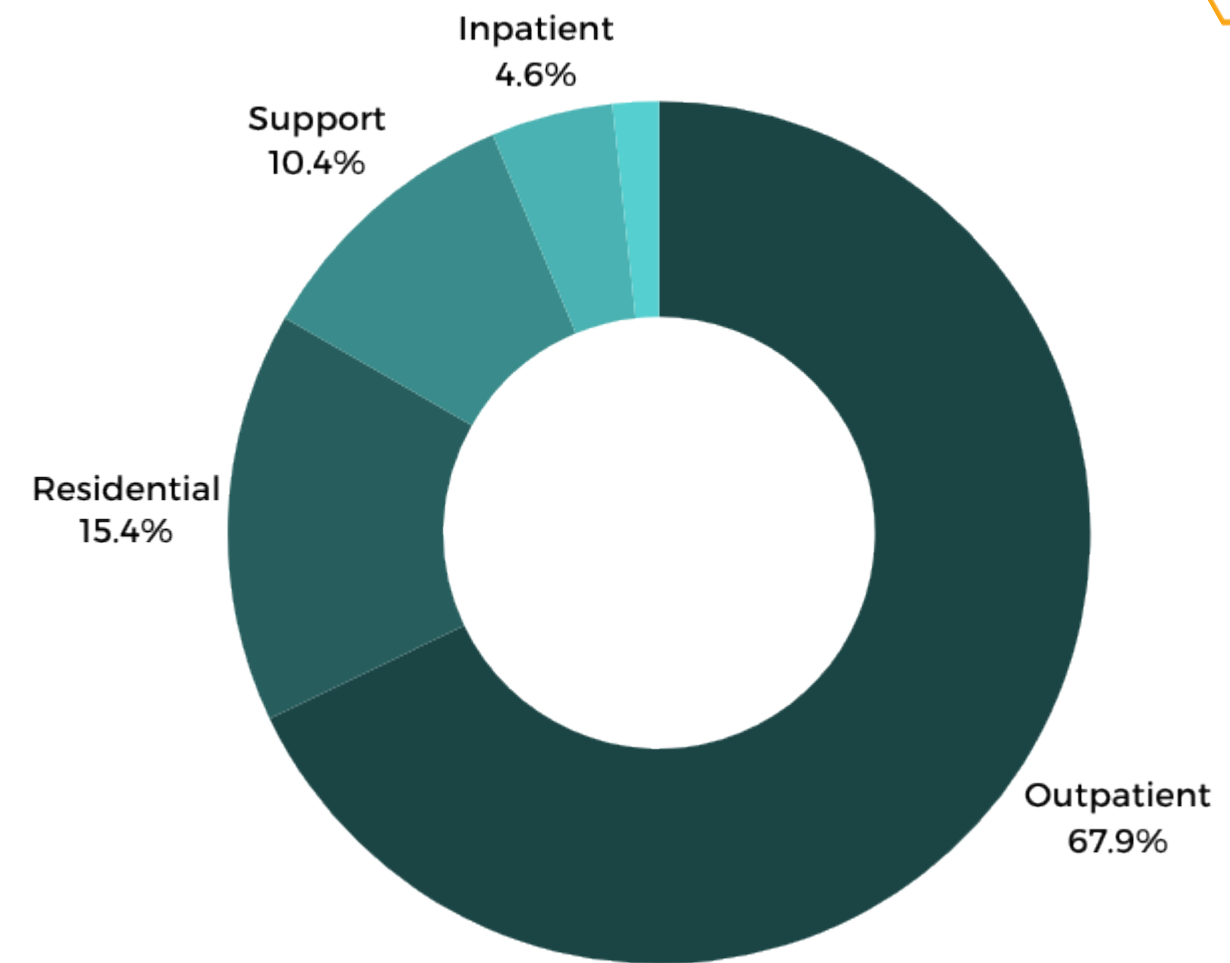


## OBJECTIVES

Revolutionize healthcare operations by deploying predictive models for proactive patient care and cost-effective service delivery.

# DATA OVERVIEW





# DATA PREPROCESSING

Data preparation included vital steps to refine and optimize the dataset for thorough analysis

Integration

Integrated external datasets using zip codes

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Cleaning

Handled missing values by eliminating rows (where infrequent), assigning new labels, and replacing unknown values.

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Correlation Analysis

Removed highly correlated columns

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Converting nominal/ordinal values to numerical values

Mapped ordinal values to numeric scale and used one-hot encoding for nominal attributes.

# METHODOLOGY



- Z-score normalization

- **K-means Clustering**

- $n = 20$

- RandomSearchCV
- GridSearchCV

- Importance values using Random Forest Classifier

- **Naive Random Oversampler**



# MODEL PERFORMANCE EVALUATION

## Datasets used

- Baseline Dataset
- Oversampled Dataset
- Oversampled PCA Dataset

## Dataset Partitioning

- 70% for training
- 30 % for testing
- Stratification

## Model Training Approach

- Trained for each dataset separately
- Evaluated using test data without oversampling

## Key Metric

- Recall

| Classification Model | Precision | Recall | Accuracy | Computation Time (in secs) |
|----------------------|-----------|--------|----------|----------------------------|
| Random Forest        | 0.98      | 0.97   | 0.97     | 215.25 secs                |
| Decision Tree        | 0.96      | 0.96   | 0.96     | 22.84 secs                 |
| Gradient Boosting    | 0.78      | 0.79   | 0.79     | 2322.39 secs               |

# PREDICTIVE INSIGHTS

Have No Insurance

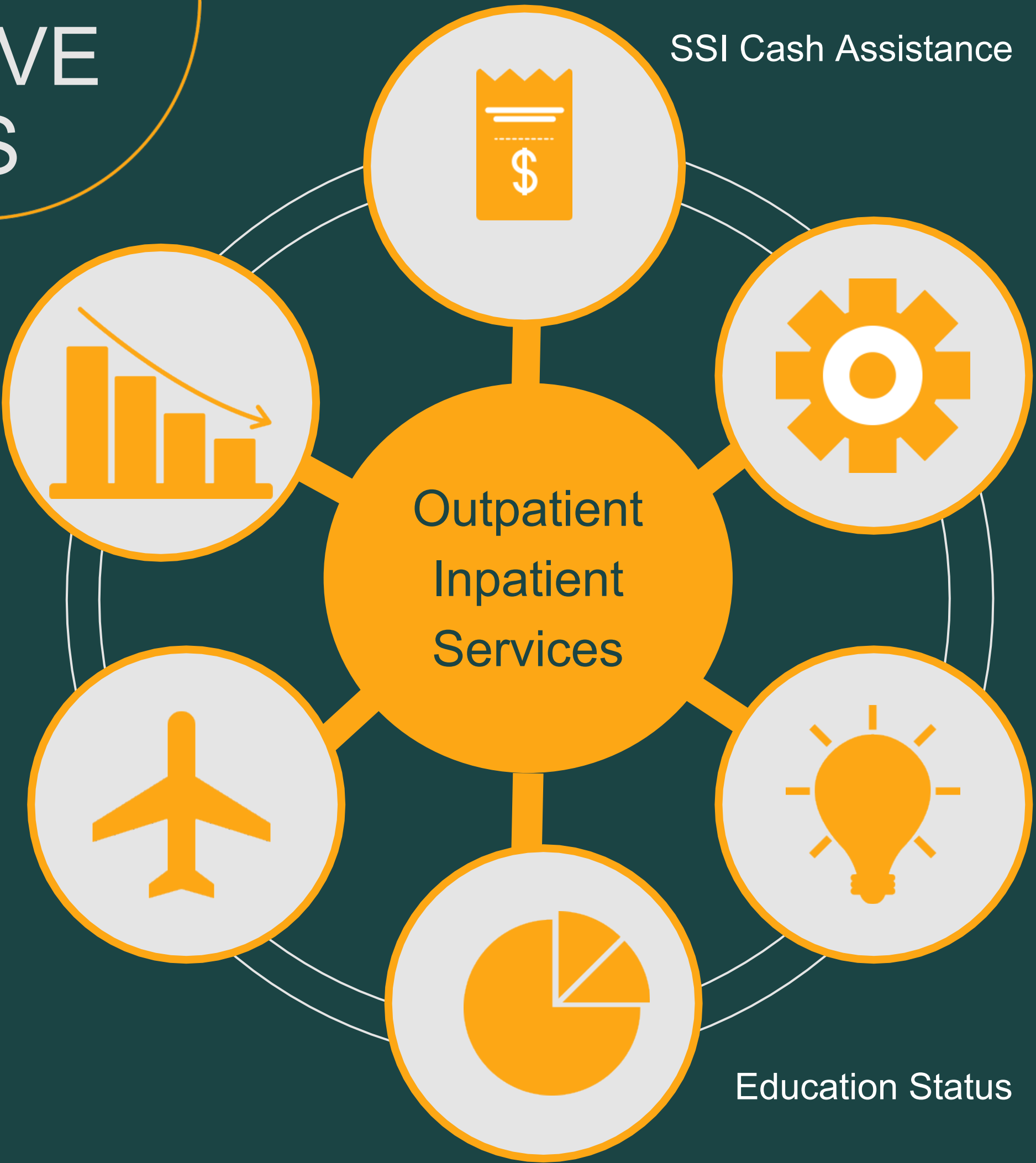
SSI Cash Assistance

Diagnosis Class: Mental Illness

From New York City

Have Hospital in vicinity

Education Status



# USE CASE AND STAKEHOLDER

## Hospital Administrators

- ◆ Resource Allocation Challenge  
Hospitals struggle to efficiently allocate resources across services, resulting in operational inefficiencies and cost concerns.
- ◆ Uncertain Service Demands  
Determining exact service needs without predictive insights leads to staffing and resource mismatches.



**Target : Hospital Administrator**  
**Use Case: Hospital Resource Optimization**

- ◆ Implement predictive models for proactive resource allocation in healthcare settings.
- ◆ Foster proactive interventions using predictive insights for regional patient care plans
- ◆ Tailor care plans based on data insights to enhance patient engagement and outcomes.

# FUTURE DIRECTIONS



Expanded Data Integration



Cross-Regional Validation



Collaborative Data Collection



Specialized Model Development



Disparity Analysis

# CONCLUSIONS



## Modelling Success and Limitations

- Successful modeling observed for non-emergency services.
- Random Forest, Decision Tree and Gradient Boosting models showcased notable performance.



## Challenges in Emergency Service Prediction

- Accurately forecasting emergency services remains a challenge due to their unpredictable nature.



## Data Validation and Model Refinement

- Pioneering advancements in healthcare service prediction models.
- Urging continuous refinement for enhanced real-world applicability.



# Thank You

Do you have any questions?