

# CUSTOMER SENTIMENT ANALYSIS IN TICKETING SYSTEM

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## Content Overview



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Google Colab
Power BI Dashboard



## DATA UNDERSTANDING



## Data Understanding

#### **About Dataset**

The Dataset is a <u>Ticket System Review</u>, it is a Dataset about a survey conducted to the customers from various ticketing platforms such as:

#### Zendesk, Zoho Desk. Freshdesk, ServiceNow, Jira Service Management, OTRS

#### **Dataset Overview**

Each survey response contains the following details:

- Survey ID & Date: A unique identifier for each survey, along with the date it was conducted.
- Ticket System: The specific platform being evaluated.
- Ratings (1-5 Scale): Includes ratings for overall experience, customer service, features, value for money, and ease of use.
- Likelihood to Recommend (NPS, 0-10 Scale): A metric used to assess customer loyalty.
- Customer Reviews: Text-based feedback from users.

## Data Understanding

#### **Objectives**

#### To Do Key Metrics Analysis

- Response Rate: Ratio of Customers who Filled the Survey
- Customer Satisfaction Score (CSAT): Overall Satisfaction of the Customers
- Customer Effort Score (CES): How much effort is needed from Customers.
- Net Promoter Score (NPS): Measure Customer Loyalty and Satisfaction
- Sentiment Analysis: Getting Insights from Text.

Additionally, We will be making a **Dashboard** to **Visualize** the Datas to make for an easier **Understanding using Power BI** and see if there are any **Trends or Patterns** within the Data



## METHODOLOGY



## Data Cleaning

- **1.**Changing Date of Survey to Date and Time Format
- 2. Making a New Dataset that are only for Customers that Responded to the Survey for Analysis

```
# Convert data type
dataset['date_of_survey'] = pd.to_datetime(dataset['date_of_survey'])
dataset.head()
```

```
# Responded Customer
responded_customer = dataset[ dataset['fill_survey'] == 'Responded'].copy()
responded_customer.head()
```

```
# Check the type of data
   dataset.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1462 entries, 0 to 1461
Data columns (total 10 columns):
                             Non-Null Count Dtype
                             1462 non-null
   id survey
    date of survey
    ticket system
                             1462 non-null
                                            object
                             787 non-null
                                            float64
    overall rating
    customer service
                                            float64
                             787 non-null
                                            float64
                             787 non-null
                                            float64
    value for money
                                            float64
    likelihood to recommend 787 non-null
                                            float64
                             787 non-null
   overall text
                                            object
dtypes: float64(6), object(4)
memory usage: 114.3+ KB
```

### **Key Metrics Calculation**

#### 1. CSAT Score

$$CSAT = rac{\Sigma \ total \ satisfaction \ score}{number \ of \ responded \ customer imes max \ rating}$$

The Customer Satisfaction (CSAT) score measures how satisfied customers are with a product or service

#### 1. NPS Score

$$NPS = \frac{Promoter - Detractor}{Total \; Survey \; Responded}$$

Net Promoter Score (NPS) is a measure customer loyalty and satisfaction based on how likely customers are to recommend a product or service to others

#### 2. CES Score

$$CES = rac{\Sigma \ total \ effort \ score}{number \ of \ responded \ customer imes max \ rating}$$

Customer Effort Score (CES) measures how much effort a customer has to put in to resolve an issue or complete a task to helps understand how easy or difficult it is for customers to interact

## Sentiment Analysis

Next, we will be performing **Sentiment Analysis**, However, in Order to get a better output in our Sentiment Analysis, we will first need to clean the text by **removing double whitespaces**, **cleaning URLs/websites**, **and eliminating usernames** (commonly found in social media or digital text).

This ensures that the Sentiment text that we will be using is clean and will be able to be processed better.

### Sentiment Analysis

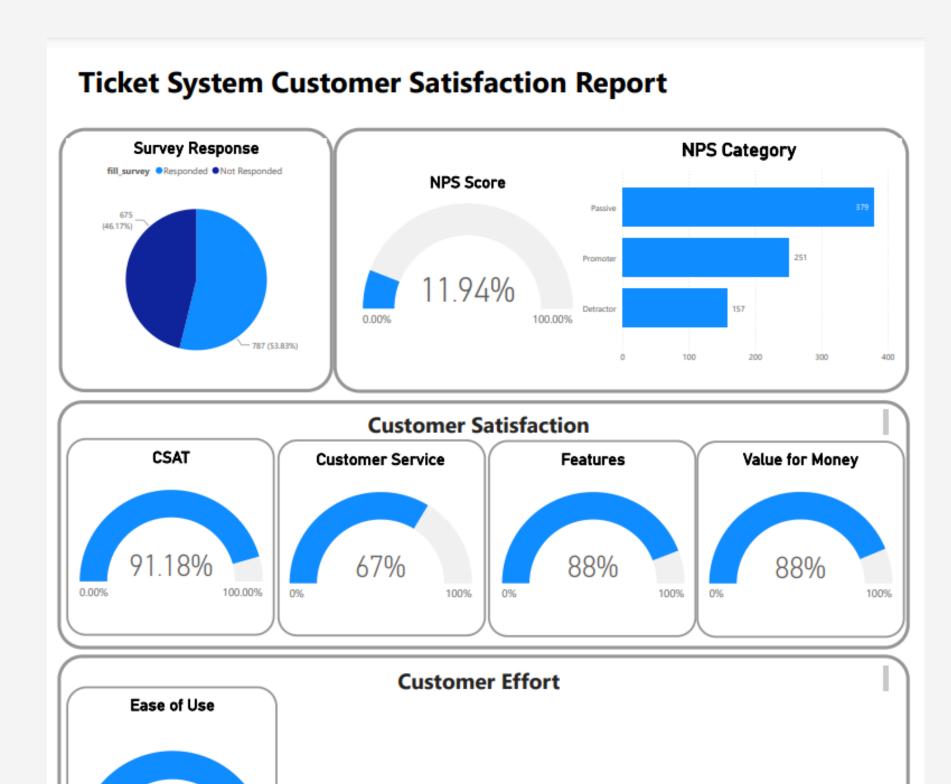
This Sentiment Analysis Model employs sentiment analysis using the Twitter RoBERTa model from Hugging Face, which is specifically trained to detect sentiment in text, especially from platforms like Twitter

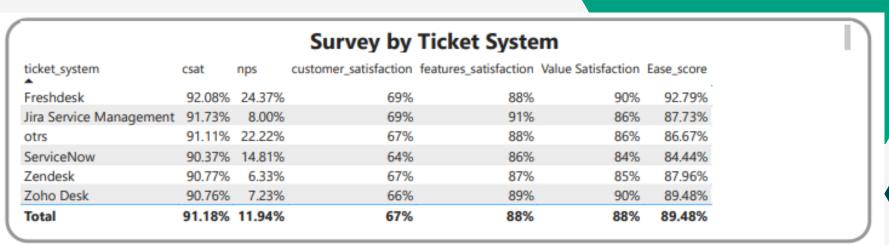
The steps involved in using the Transformers library with the sentiment-analysis pipeline include:

- Classifying text into three categories: Negative (LABEL\_0), Neutral (LABEL\_1), and Positive (LABEL\_2).
- Storing the analysis results in the sentiment\_score and sentiment columns.

This methodology is implemented end-to-end, from basic data cleaning to advanced analysis, utilizing machine learning models and major Python libraries such as Pandas for data manipulation and Transformers for natural language processing.

### Dashboard

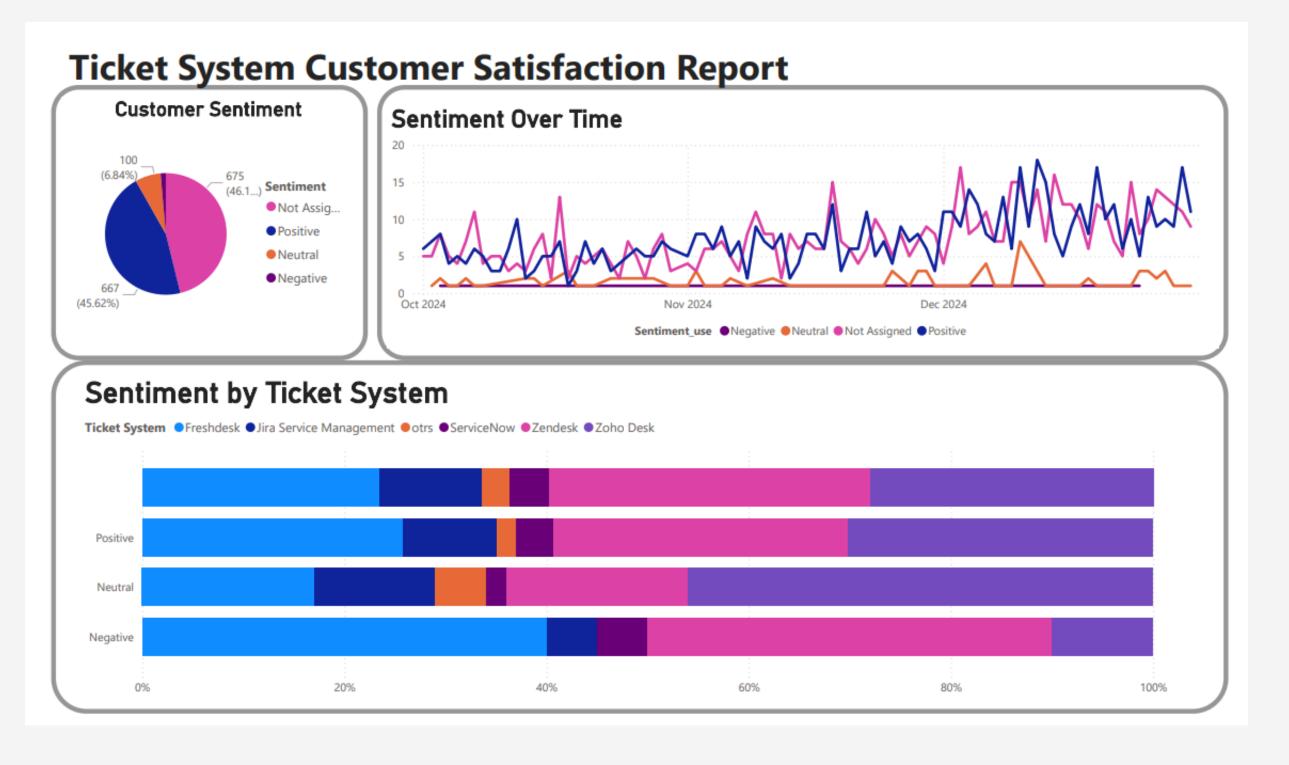






## **Ticket System Customer Satisfaction Report Dashboard**

#### **Ticket System Customer Satisfaction Report Dashboard**





## INSIGHT & RECOMMENDATION



## Insight & Recommendation



**The NPS Score** is quite low at 11.9%, While it's positive, it's not extremely high, suggesting that there's some room for improvement in customer loyalty and the Customers aren't likely to Recommend the Service to Others.



**The CSAT Score** at 91.18% shows that the overall satisfaction of the respondents is excellent. The high CSAT is a good indicator that the company is delivering a satisfactory experience to most of its customers.



**Customer Satisfaction** is high overall, with positive feedback on features (88%) and value for money (88%). However, customer service needs improvement, as it received the lowest satisfaction score of 67%, indicating potential areas for enhancement in support and responsiveness



**Customer Effort Score** at 89.48% suggests that most customers find the product easy to use, with minimal effort required to navigate or operate the system. indicating that the user experience is generally smooth and intuitive, contributing positively to customer satisfaction

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## Insight & Recommendation



Freshdesk has the highest NPS score and ease of use, setting a strong benchmark for other ticketing systems to learn from and improve their customer experience and user interface design to see what works best.

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#### Recommendation



Focus on enhancing customer service by addressing pain points, gathering feedback from detractors to understand their concerns, and continuously improving the product. Additionally, incentivize referrals and optimize the customer onboarding process to increase satisfaction and loyalty.



Maintain Customer Satisfaction, continue delivering a positive customer experience by ensuring consistent product quality and support. Monitor feedback to identify any emerging issues and address them proactively.



Improve Customer Service, focus on training and upskilling support staff to enhance their responsiveness and problemsolving capabilities. Additionally, consider implementing a more efficient support system, such as live chat or self-service options, to address customer issues faster and increase overall satisfaction.



For **Survey Response Rate**, consider implementing strategies like sending reminder emails to customers who haven't completed the survey, offering incentives for participation, making the survey shorter and easier to complete. also ensure that the survey is easily accessible across different devices and platforms to encourage more customers to respond

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### Thank You!

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