



# OKLAHOMA CITY UNIVERSITY

## MASTERS IN COMPUTER SCIENCE

PROJECT  
ON

# TOUR SALES PREDICTION USING MACHINE LEARNING

UNDER THE GUIDANCE OF  
TASHFEEN, AHMAD  
MATHEMATICS & COMPUTER SCIENCE

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# Overview:

This project focuses on building an intelligent web-based system to predict whether a customer will accept a tour package. The system leverages historical customer data to:

- Analyze behavioral and demographic trends
- Train a machine learning model to identify patterns
- Allow real-time predictions through a user-friendly interface built using Flask

## Purpose:

To assist tour operators and marketers in targeted decision-making by identifying potential customers likely to buy a tour product, thereby improving sales strategies and resource allocation.



**Predicting Customer Acceptance of Tour Packages**  
**Enhancing Sales Strategies with Customer Behavior Analysis**



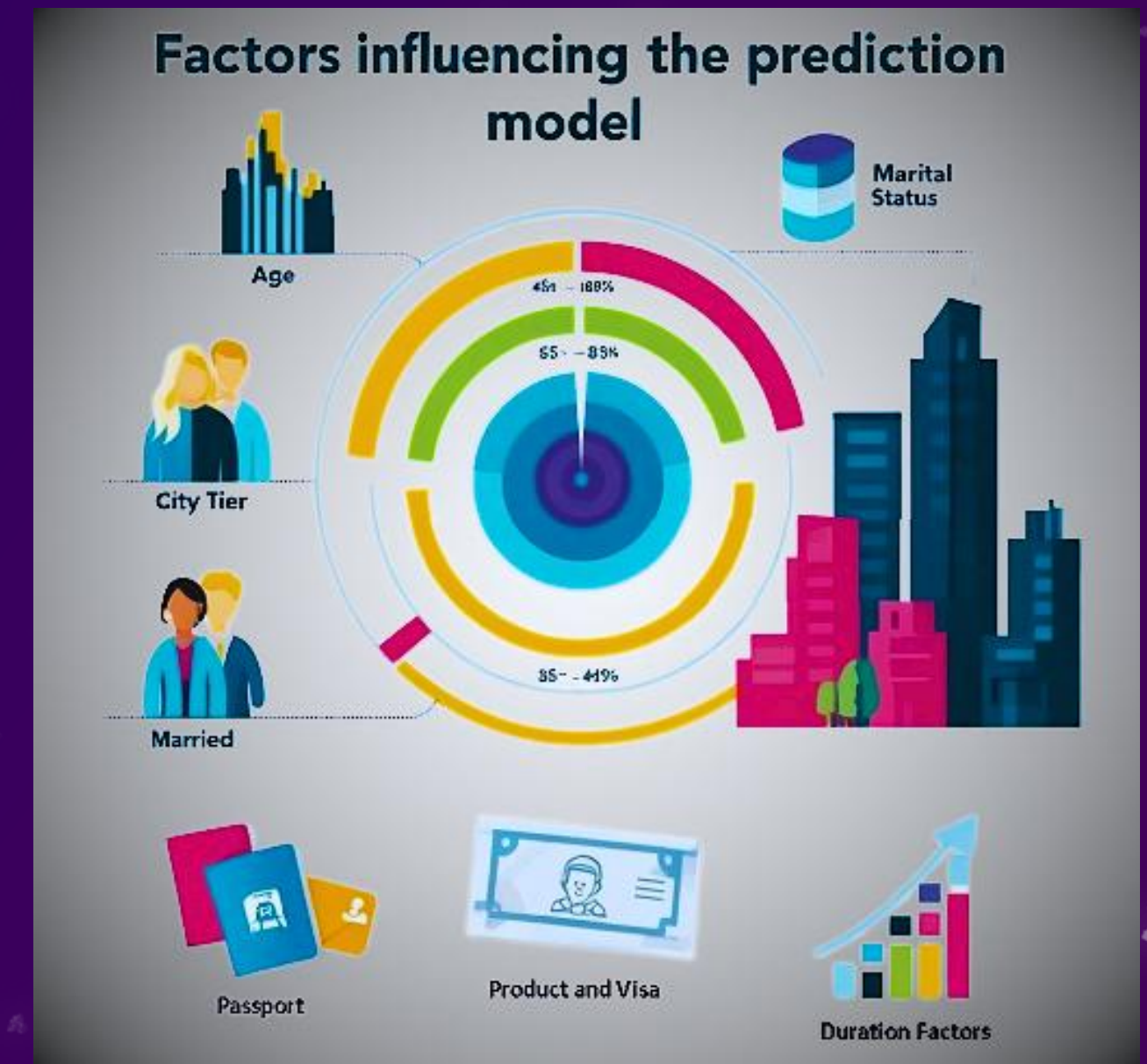
# Background – Problem Statement, Data & Approach

## Problem Statement:

Tourism companies face challenges in understanding customer interests and behaviors. With numerous variables like age, occupation, marital status, travel history, and passport details, manual prediction is inefficient and inaccurate.

## Dataset Details:

- Contains features such as:
  - Age, City Tier, Marital Status, Passport, Visa, Product Pitched, Duration, etc.
  - Target column: ProdTaken (1 = Product Taken, 0 = Not Taken)
- Collected from a travel company's customer records



# Model Training – Findings & Process

## Training Process:

- Features like Age, Passport validity, City Tier, and Duration used to train the model
- Preprocessed using label encoding and scaling (if needed)
- Split into train-test (80/20 ratio)
- Model trained using RandomForestRegressor( $n\_estimators=100$ )

## Key Metrics:

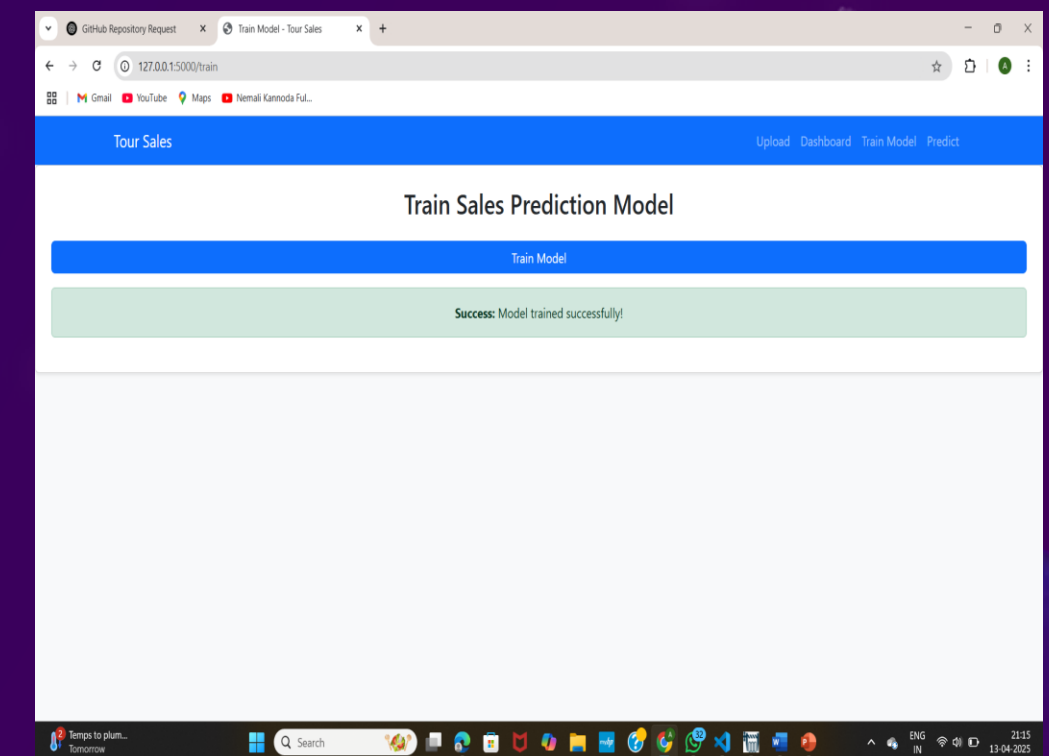
- Mean Absolute Error (MAE): Measures average prediction error
- Mean Squared Error (MSE): Penalizes larger errors
- $R^2$  Score: Measures how well predictions match actual outcomes

## Interpretation:

- High  $R^2$  indicates excellent model fit
- Low MAE/MSE indicates minimal deviation from actual values

## Conclusion:

The model is highly accurate in identifying customers likely to purchase the tour package.





# Operation of the Model – Real-Time Prediction Flow

## User Interaction Flow:

1. Upload CSV File – Raw data file is uploaded via web UI
2. Data Processing – Missing values handled, categorical features encoded
3. Visualization – User selects any column to visualize trends (e.g., Bar chart)
4. Training Triggered – User clicks 'Train' and model is trained in the background
5. Prediction Interface – User inputs individual customer details via a form
6. Prediction Result – Backend loads trained model and returns prediction instantly

## Key Implementation Components:

- Flask routing (/upload, /dashboard, /train, /predict)
- pickle to load/save trained model
- session to manage user data through the app

## Benefit:

Real-time predictions help sales teams personalize outreach and increase conversion rates.

The screenshot shows a web browser window with the URL '127.0.0.1:5000/predict'. The page title is 'Predict Customer Conversion'. The form contains the following fields and labels:

- Age:
- City Tier:
- Number of Follow-ups:
- Preferred Property Star:
- Pitch Satisfaction Score:

Below the form is a green button labeled 'Predict'. Underneath the button, a pink box contains the text: 'Prediction: Customer is unlikely to buy the tour package.'

# Conclusion & Acknowledgements

## Key Achievements:

- Built an end-to-end intelligent web application
- Achieved high prediction accuracy ( $R^2 > 0.90$ )
- Integrated data processing, visual analytics, and ML deployment
- Provides actionable insights to tour companies

## Acknowledgements:

- Tools & Libraries: Python, Flask, Pandas, Scikit-learn, Matplotlib
- Special thanks to project mentors, dataset providers, and academic support



**THANK YOU**