

UK Train

Data Analysis Project



Problem Statement:

- The UK train system faces continuous challenges such as unpredictable delays, fluctuating passenger demand, unclear ticket choices, and revenue uncertainty.
- Passengers struggle to select the best ticket option, while decision-makers lack a unified tool to analyze performance, understand trends, and predict future behaviors.

Proposed Solution:

- We developed an intelligent, data-driven dashboard that analyzes train journeys, predicts operational risks, and provides smart ticket recommendations.
- The system combines traditional analytics with Machine Learning to offer:
- Passenger demand forecasting: predict the volume of passengers per month
- Journey performance prediction: predict cancelled journeys per month
- Operational risk analysis: predict average delay time per month
- Delay prediction and risk scoring
- All integrated into an interactive, user-friendly dashboard that helps both passengers and decision-makers make informed choices.

Unique Value Proposition:

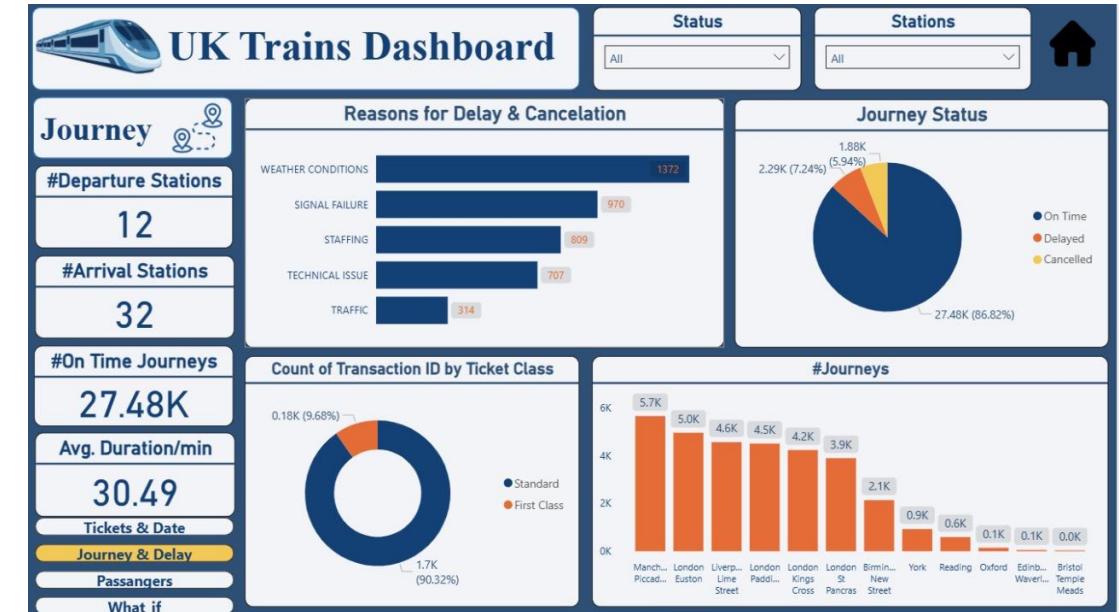
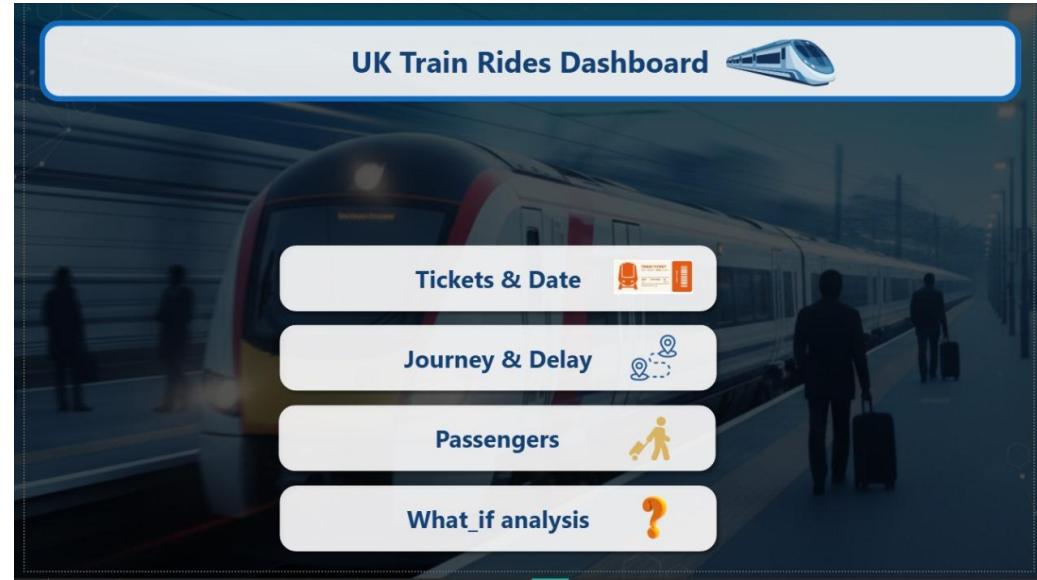
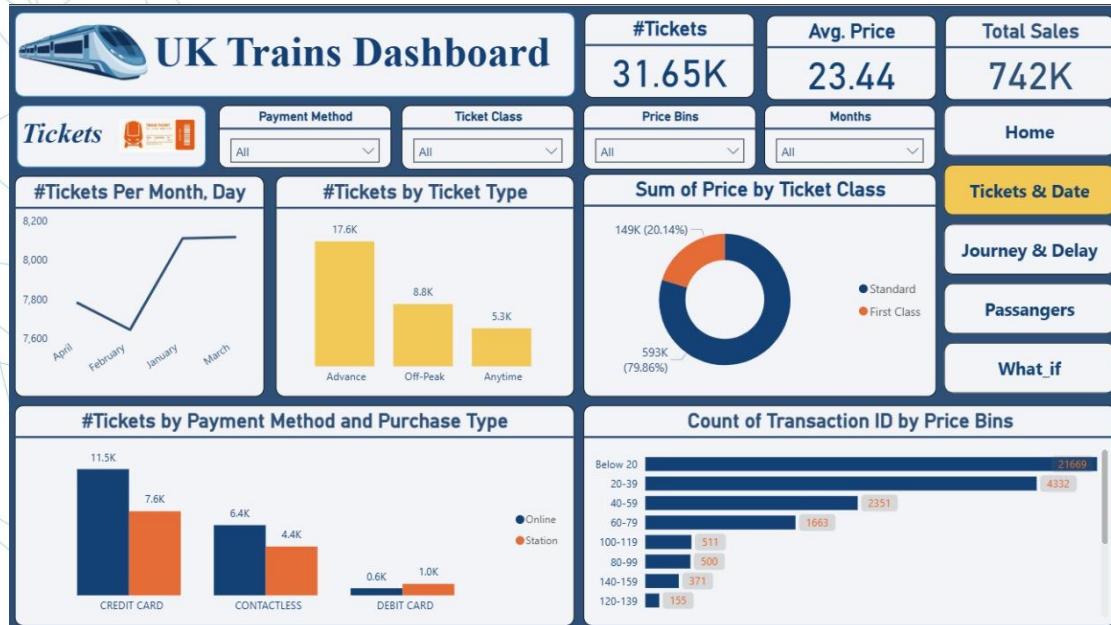
Our project stands out by transforming raw railway data into a clear, interactive, and insight-driven dashboard.

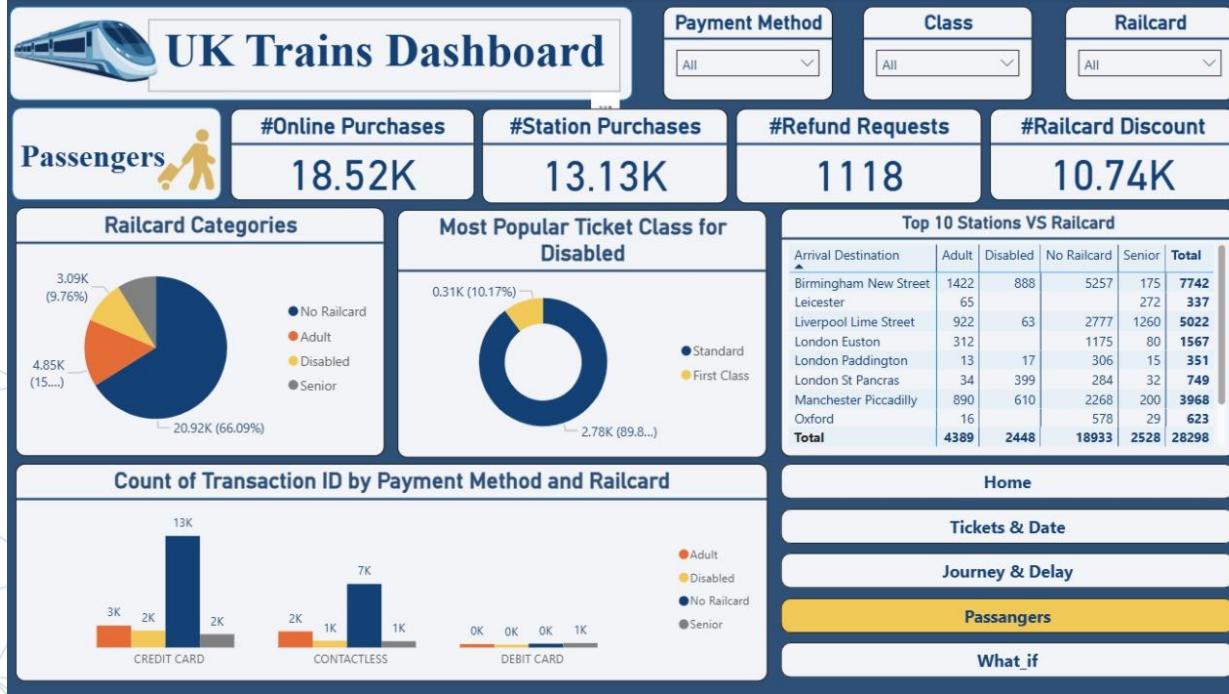
Unlike regular reports, our solution offers:

- Comprehensive trend analysis for passenger demand and journey activity, including predictions for the volume of passengers per month
- Detailed delay insights using fields such as Journey Status, Actual Arrival Time, and Journey Delay (min), alongside predicted average delay time per month
- Forecasting of cancelled journeys per month to help anticipate operational disruptions
- Route performance comparison across different departure and arrival stations
- Ticket analysis by Ticket Class, Ticket Type, Railcard, and Purchase Method
- Revenue and pricing breakdowns supported by Price, Price Bins, and booking times
- A structured star-schema data model that improves accuracy, flexibility, and reporting performance

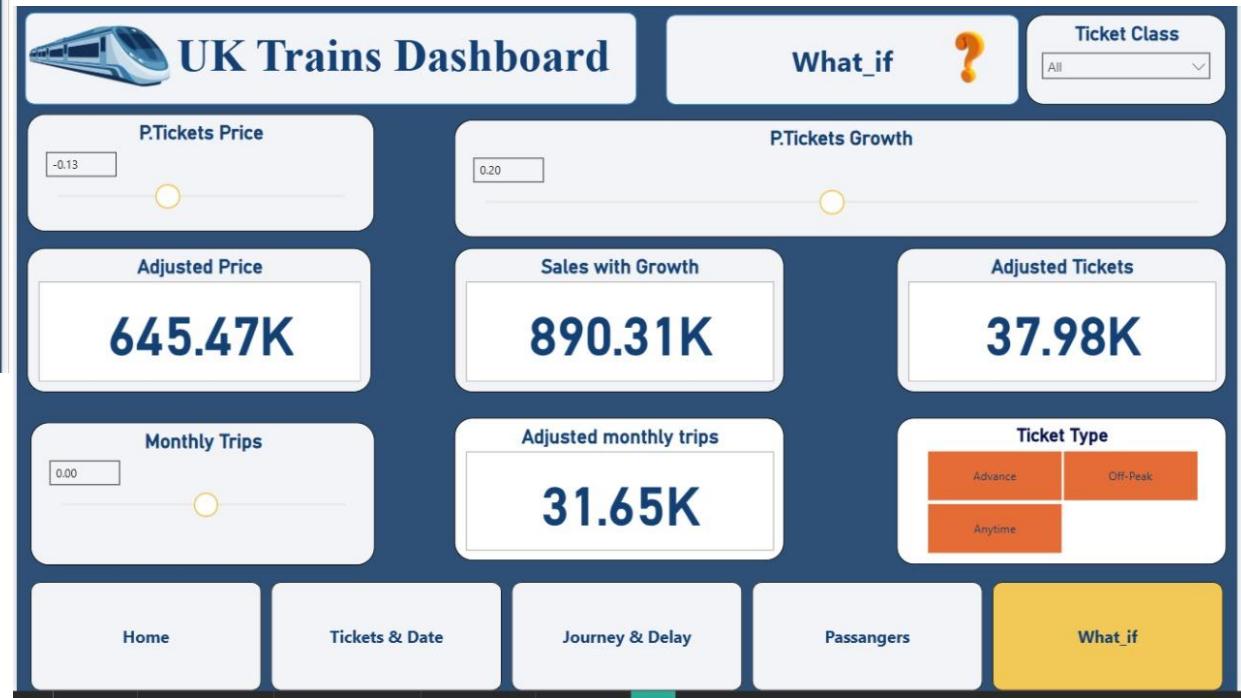
This makes the system not just a visual dashboard, but a practical tool that helps understand travel patterns, optimize operations, and support data-driven decision-making with predictive insights.

Dashboard:





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End Users + Features:

- Passengers / Travelers: Need to select the best ticket, avoid delays, and plan trips efficiently.
- Railway Operators / Businesses: Monitor journey performance, passenger volumes, revenues, and optimize operations.
- Decision-makers / Planners: Require accurate forecasts for passenger demand, cancellations, delays, and financial insights to make strategic decisions.

Key Features & Insights:

- Passenger Demand Forecasting → Predict monthly passenger volume
- Delay & Risk Prediction → Estimate journey delays and average delay times
- Cancelled Journey Forecasting → Forecast number of cancelled journeys
- Revenue & Pricing Analysis → Analyze revenue and prices by Ticket Class and Payment Method

How Features Solve User Problems:

User Persona	Problem	Feature/ Insight	Benefit
Passengers	Difficulty selecting the best ticket	Ticket Recommendation, Ticket Analysis	Choose the best ticket, save money and time
Passengers	Uncertainty about delays or cancellations	Delay & Cancel Forecasting	Plan trips better, avoid risky journeys
Operators	Hard to monitor operations and resources	Passenger Demand Forecasting, Revenue Analysis	Optimize staffing, train allocation, and sales strategies
Decision-Makers	Need accurate operational predictions	Predictive Insights (Delay, Cancel, Passenger Volume)	Make strategic, data-driven decisions

Data Structure:

Database Architecture:

- Data stored in CSV files, representing structured tabular format.
- Flat file structure, no relational database used, easy to load into Python/Power BI.

Key Entities & Features:

- Tickets: Transaction ID, Price, Ticket Class/Type, Railcard, Payment Method, Purchase Date
- Passengers: Passenger categories (Adult, Senior, Disabled, No Railcard), Count per ticket type
- Journey: Departure & Arrival Stations, Journey Status, Actual Arrival Time, Delay (min), Reasons for Delay



Data Sape & Quality:

- Total rows: +30k Row
- Columns: 32 features including categorical, numerical, and date/time fields
- Data cleaning applied: removed duplicates, fixed date/time formats, handled missing values, converted times to numeric, encoded categorical values
- Data validation performed to ensure consistency and correctness
- Some imbalance present (e.g., delayed vs. on-time journeys), considered in predictive models

Data Flow:

- Collection: Train booking systems export CSV files
- Storage: Stored locally or on server as CSV
- Processing: Loaded into Python/Power BI for analysis and predictive modeling
- Access: Filtered and visualized in the dashboard; used for forecasting and AI predictions

Programming Languages & Frameworks

Main Languages:

Python (data preprocessing, ML models, forecasting)

Frameworks & Tools:

- Pandas, NumPy (data manipulation and analysis)
- Scikit-learn (machine learning models: Random Forest, Logistic Regression)
- Matplotlib, Seaborn (data visualization)

Supporting Technologies:

- CSV files as primary data source
- Power BI (visualization, dashboard deployment)

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Final Project Deliverables

- Fully interactive UK Trains Analytics Dashboard (Power BI)
- Cleaned & validated CSV dataset (after removing duplicates, fixing dates, and applying validation rules)
- Machine Learning notebooks for:
 1. Passenger demand forecasting
 2. Cancelled journeys prediction
 3. Average delay time prediction
- Python script for data preprocessing

Documentation Provided

- Data Cleaning & Preparation Document
- Exploratory Data Analysis (EDA) Report
- Forecasting
- Power BI Dashboard Guide
- Testing & Validation Report
- Project Summary + Conclusions



Source Code & Files

- Power BI project file (.pbix)
- GitHub repository including:
 1. Notebooks
 2. Python script
 3. Code of forecasting
 4. Original Data
 5. Cleaned Data
 6. Dashboard file

Timeline & Milestones

- Week 1–2: Data cleaning + validation
- Week 3: Data modeling + dashboard draft
- Week 4: ML modeling & forecasting
- Week 5: Integration + testing
- Week 6: Final dashboard + documentation delivery

Team Members & Their Key Responsibilities:

- **Marina Youssef:** Data Cleaning(Python), Forecasting(Python), Visualization(Power BI), Deployment(Streamlit), Analysis(Python)
- **Helana Hany:** Data Cleaning(Python), Forecasting(Python), Analysis(Python)
- **Salma Hazem:** Data Cleaning(Python), Forecasting(Python), Visualization(Power BI), Deployment(Streamlit), Analysis(Python)
- **Mariam Ahmed:** Data Cleaning(Python), Forecasting(Python), Visualization(Power BI), Analysis(Python)
- **Menna-Allah Mahmoud:** Data Cleaning(Python), Analysis(Python)

Team Members & Roles:

- **Marina Youssef:** Data Analysis, Visualization, Forecasting
- **Helana Hany:** Data Analysis, Forecasting
- **Salma Hazem:** Data Analysis, Visualization, Forecasting
- **Mariam Ahmed:** Data Analysis, Visualization, Forecasting
- **Menna-Allah Mahmoud:** Data Analysis

Collaborations Methods:

- Communication: Zoom / WhatsApp / Google meeting
- Code Sharing: GitHub Repository
- Project Management: Agile approach (sprints, task assignment)
- Version Control: Git + GitHub branches

Thank You!

[GitHub Repository](#)

[Deployment Link](#)

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