



EXECUTIVE SUMMARY



Objective: Analyze SpaceX launch data. Why: Reduce costs by improving booster recovery rates. Data: SpaceX API, mission records.



Collected from SpaceX API, cleaned missing values, corrected data types, engineered features like binary landing outcome.



Used Matplotlib, Seaborn, Plotly for scatterplots, booster success, launch trends. Built Plotly Dash dashboard for real-time filtering.



Binary classification using Logistic Regression. Features: Payload, orbit, booster. Target: Landing success. Evaluated accuracy & recall.



Key findings: Heavier payloads decrease success; FT boosters perform better; CCAFS site has higher success. Visualized with charts, scatterplots.



Used SQLite: counted missions/site, avg payload/orbit, booster success rates. SQL JOINs, GROUP BY, aggregates revealed hidden trends.



Folium maps: marked launch sites, visualized distances, highlighted geographic factors affecting landings.



Interactive dashboard with filters for site, booster, outcome. Graphs update dynamically, trends explored without code.



Model accuracy ~80%. Key features: Payload, booster, site. Provided confusion matrix, classification report, feature importance.



Summary: Combined wrangling, SQL, EDA, geospatial analysis, ML to predict SpaceX landings. Future: more features, ensemble models, API.