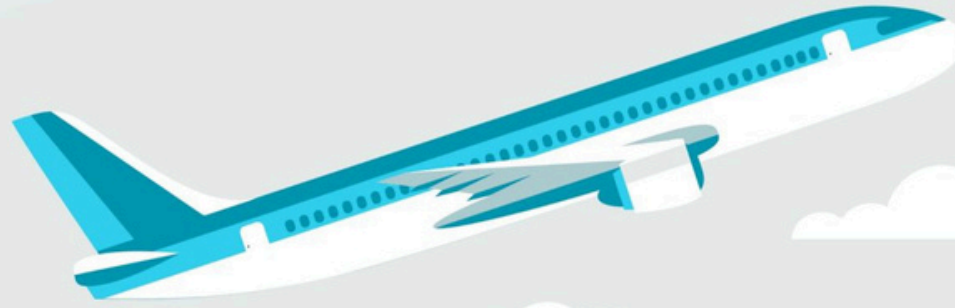


American  
Airlines



 DELTA



# FLIGHT PATH

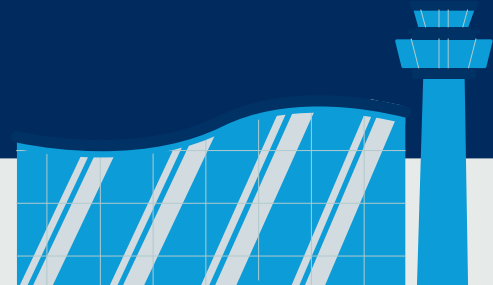
Flight Delay and Customer Experience Analysis for Operational  
Efficiency and Satisfaction





# FLIGHT PATH

Deployment of Flight Delay  
Prediction System



# AGENDA



- About the Project
- Feature Selection
- Model Development and Evaluation
- Gradio Interface Overview
- Deployment Workflow
- Results and Insights

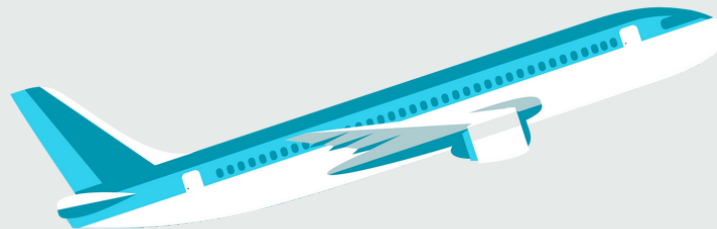
# About the Project



**Brief overview of the goal:** Predicting flight delays for American Airlines and Delta.

**Objective:** Minimize delays by providing actionable insights.

**Key Focus:** Machine learning-based prediction system deployed via Gradio.

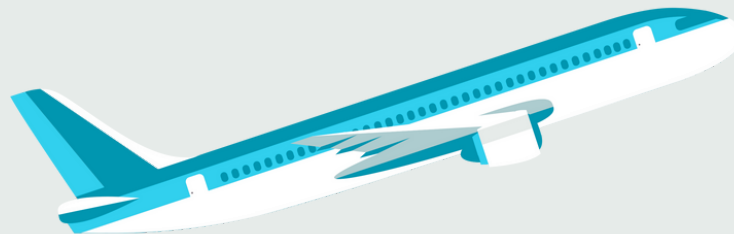


# Data Overview



**Data source:** Flight delay data extracted from Bureau of Transportation and Statistics, U.S. From **2014 to 2023**.

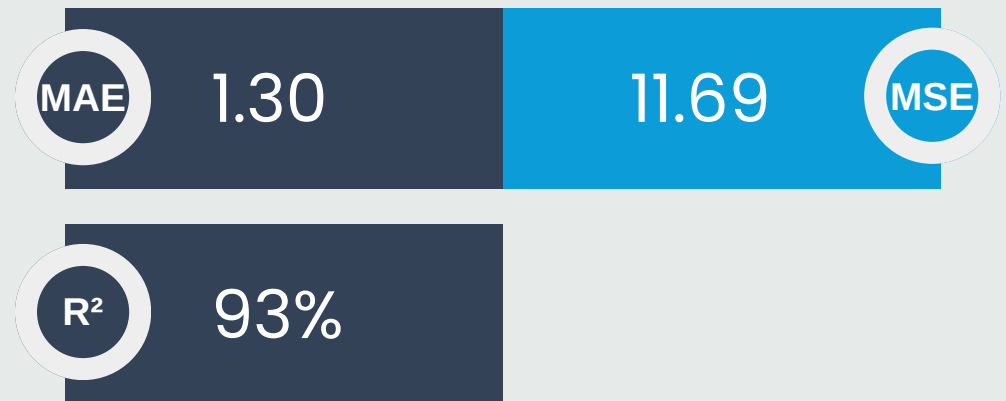
**Comparison:** American Airlines and Delta in 5 major US airports (ATL, ORD, DFW, DEN, LAX).



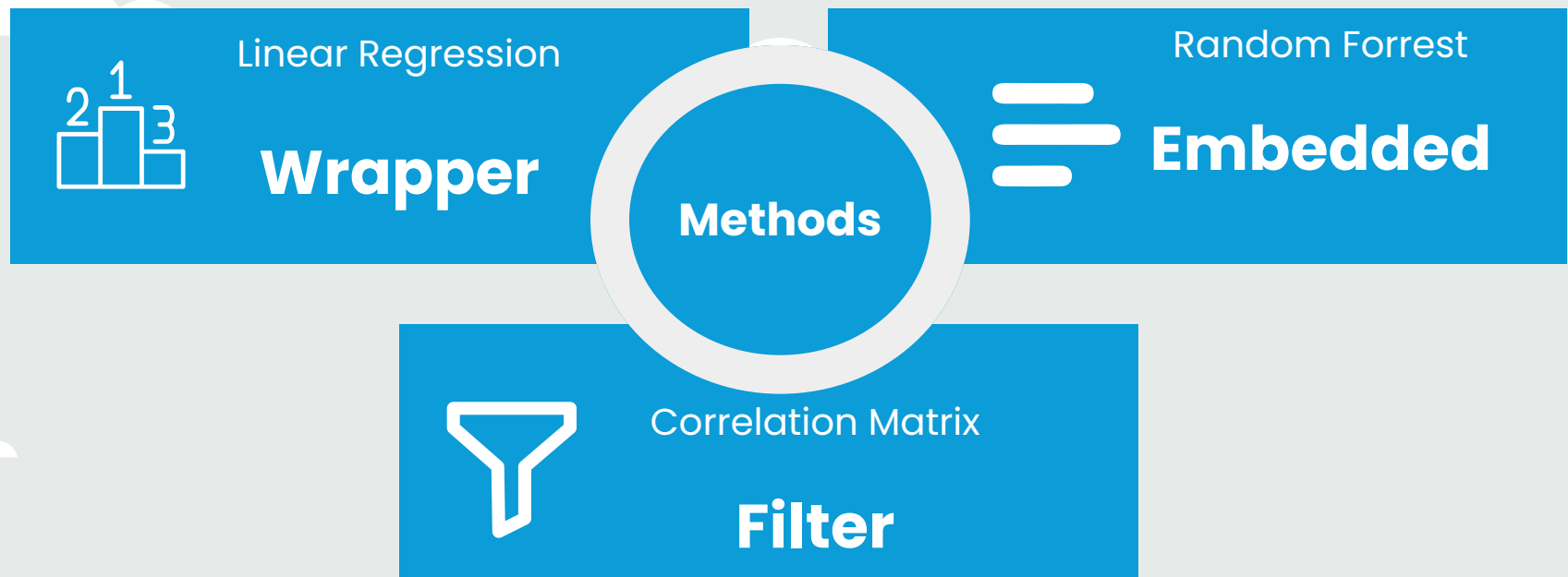
# Baseline Model



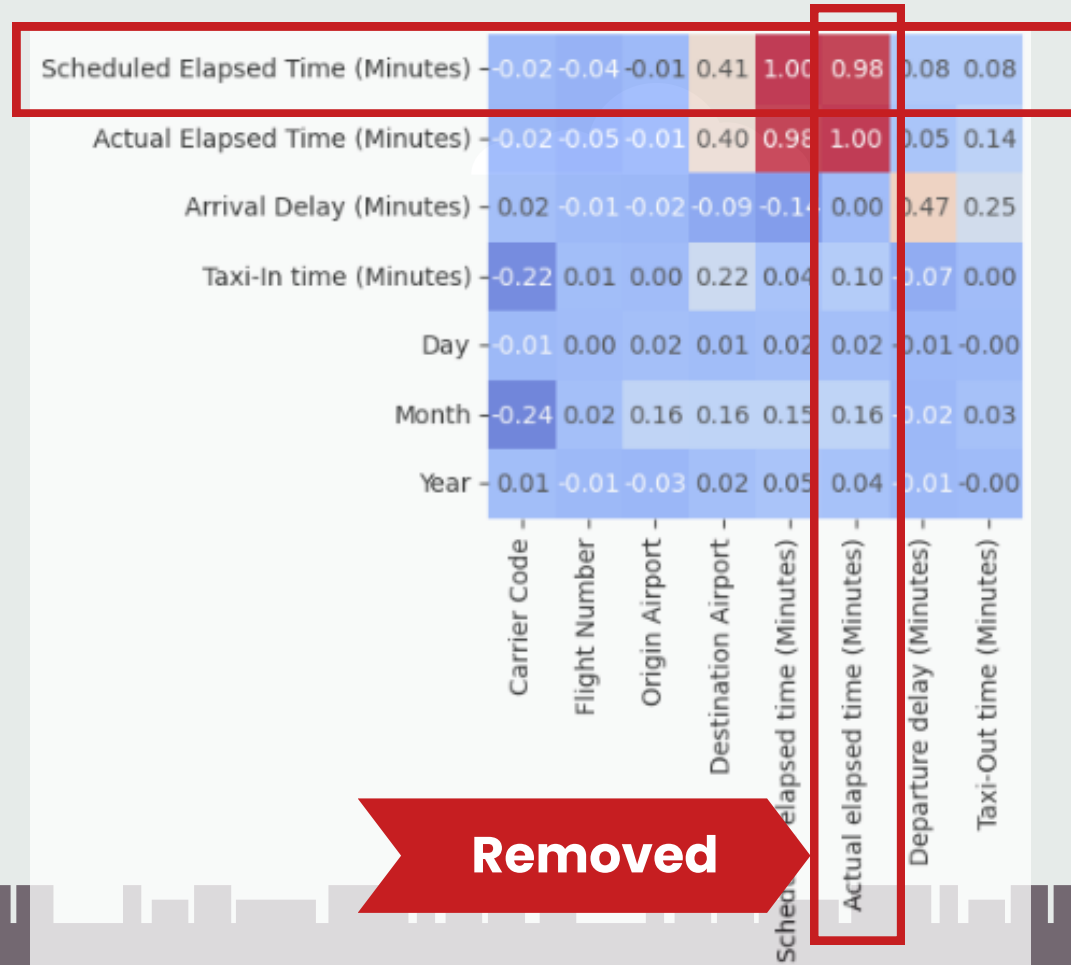
- Feature Engineering and Selecting relevant features for regression.
- Train a Regression Model (RF).
- Evaluate the Model.



# Feature Selection ↘



# Feature Selection





# Selected Features



**Date-related:** "Day",  
"Month", "Year".

**Operational Times:** Taxi-Out time,  
Taxi-In time, Scheduled Elapsed Time.

**Delay & Schedule:** Departure delay, Scheduled  
departure hour.

**Categorical:** Origin Airport,  
Carrier Code.

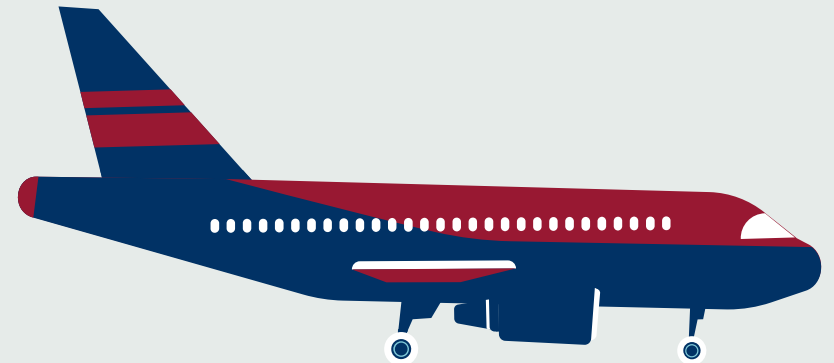


# Model Overview



## Final Models Used:

- **MLP Neural Network** (deep learning approach).
- **XGBoost** (traditional tree-based method).
- **Approach:** Multiple Models run in Pycaret.

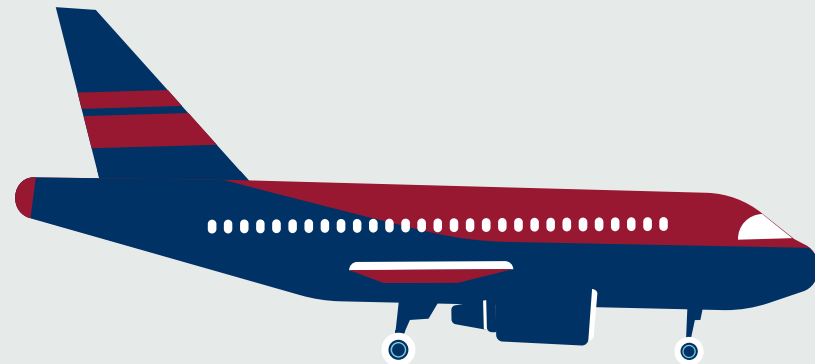




	Model	MAE	MSE	RMSE	R2	RMSLE	MAPE
<b>xgboost</b>	Extreme Gradient Boosting	5.2643	52.3456	7.2348	0.5792	0.6705	0.8301
<b>lightgbm</b>	Light Gradient Boosting Machine	5.5776	57.0497	7.5529	0.5414	0.6945	0.8660
<b>rf</b>	Random Forest Regressor	5.7253	59.7534	7.7298	0.5197	0.7107	0.8855
<b>et</b>	Extra Trees Regressor	5.7157	59.8784	7.7378	0.5187	0.7127	0.8857
<b>gbr</b>	Gradient Boosting Regressor	6.1442	67.2644	8.2013	0.4593	0.7275	0.9586

# MLP Model ↘

- **Architecture:** Multi-layer perceptron with several hidden layers.
- **Large dataset:** With 200K+ records, deep learning models benefit from more data and generalize better.
- **Optimized:** Regression tasks.



# MLP Model Performance



# Comparing Model Performance



**MLP**

100 epochs

Train MAE

**5.11**

Test MAE

**5.17**

$R^2$  Train

**0.63**

$R^2$  Test

**0.63**

**XGBoost**

Train MAE

**3.67**

Test MAE

**4.9**

$R^2$  Train

**0.79**

$R^2$  Test

**0.63**

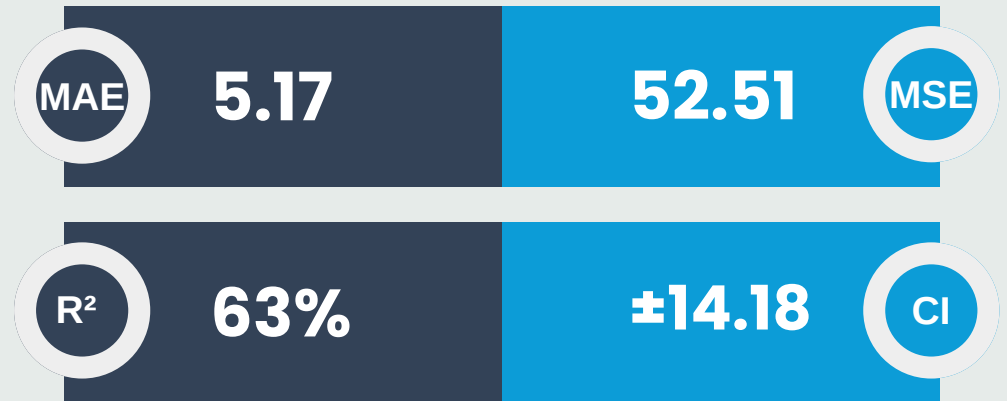
# MLP Model Evaluation



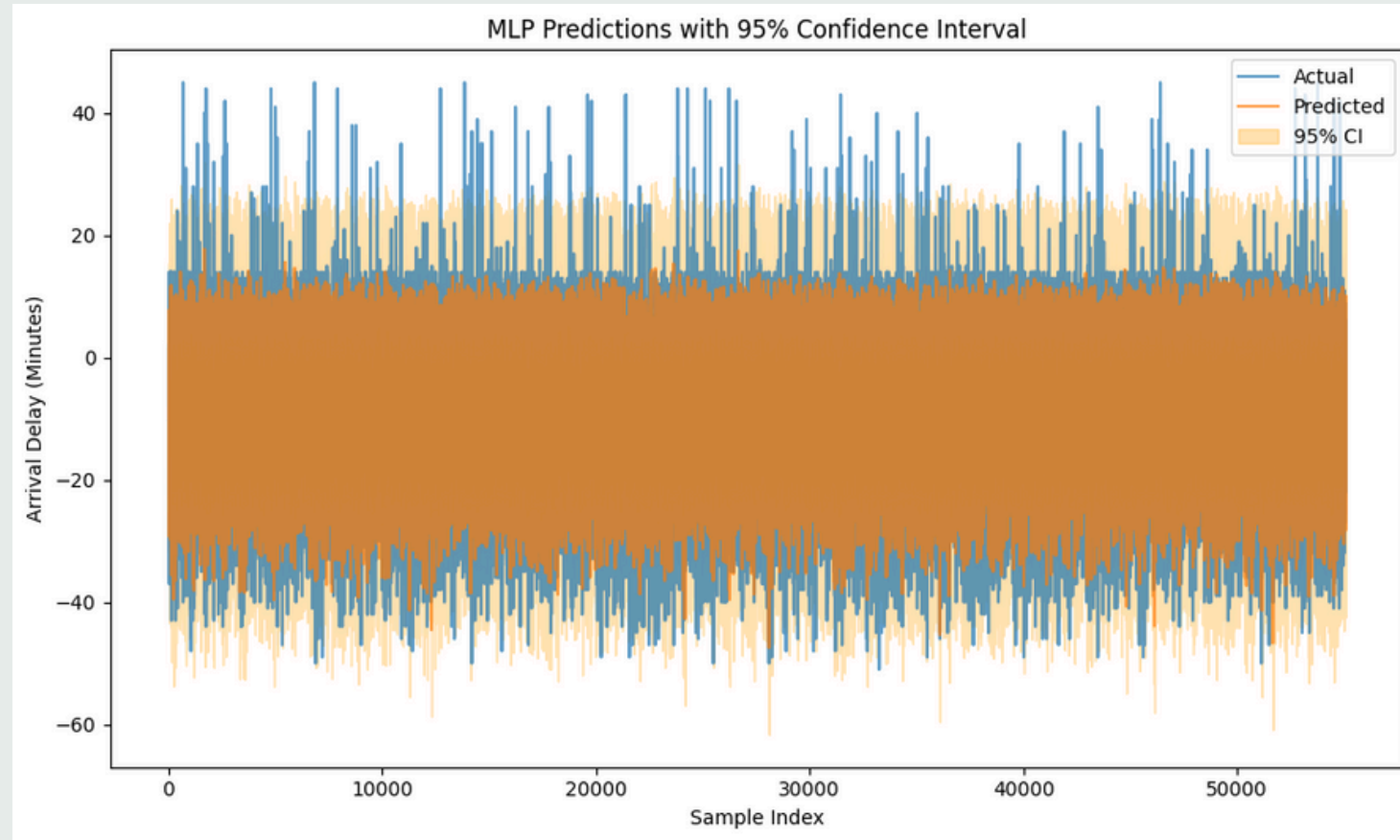
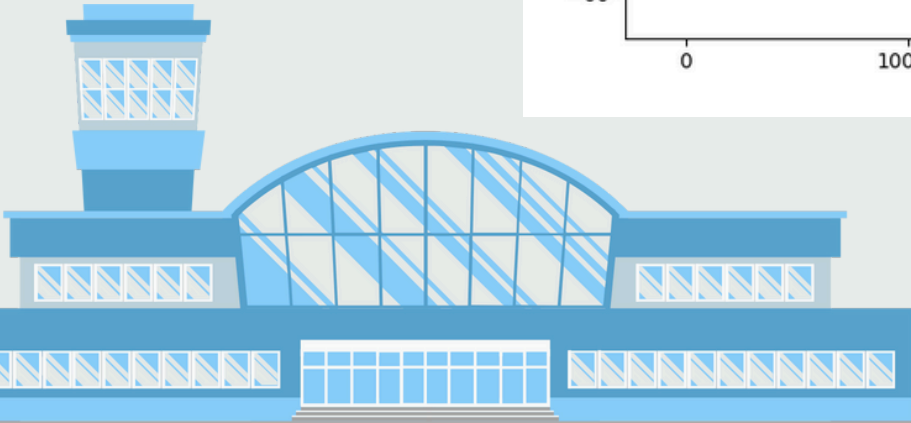
Consistent Generalization (Not Overfitting)

The train and test metrics are very close, especially  $R^2$  (both  $\sim 0.63$ ) and MAE.

Predictions are accurate within the **95% Confidence Interval** of  $\pm 14.18$  min



# 95% Confidence Interval





# Deployment Goals



Deliver a **user-friendly** interface for non-technical users.

Provide **real-time** predictions of flight arrival delays.

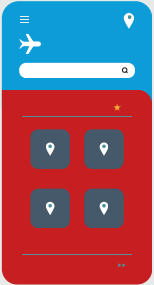
Offer **actionable insights** and recommendations for airlines.



# Gradio Workflow

User enters flight details

1



System process data and apply models

2

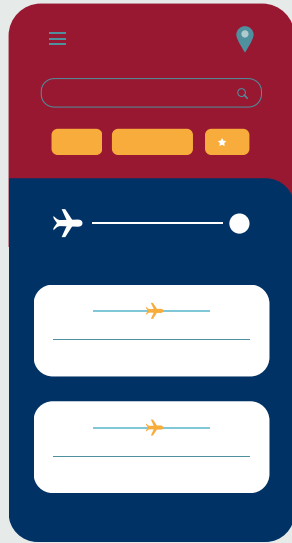
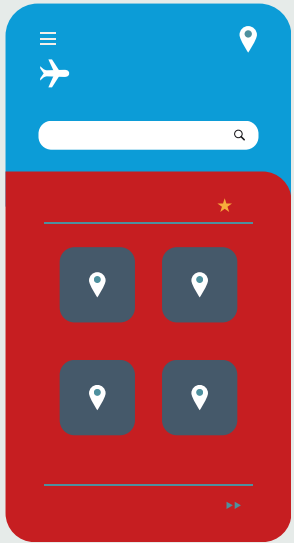


Predictions and Insights displayed in real-time

3



# Gradio Interface Overview ↘



Web-based application to interact with the models.

## Input Fields:

- Flight date, taxi-out time, taxi-in time, etc.
- Dropdowns for selecting airport and carrier.

## Output Fields:

- Predicted arrival delay, best and worst reviews.
- AI-generated operational recommendations.



# Gradio Workflow



User enters flight details



## Flight Information

Day

1

Month

2

Year

2025

## Delay & Time Information

Taxi-Out Time (Minutes)

10

Taxi-In Time (Minutes)

5

Scheduled Elapsed Time (Minutes)

130

Scheduled Departure Hour

14

Departure Delay (Minutes)

15

## Flight Route Information

Origin Airport

ATL

Carrier Code

AA


Destination Airport

DFW





# Gradio Workflow


System processes the data, applies the models


 Predict Arrival Delay

Predictions and insights are displayed in real-time

 Predictive Arrival Delay (Minutes)

8 Min 10 s 

 Based on similar past flights, this prediction is expected to be accurate within  **$\pm 14.17$  minutes** most of the time (95% confidence).

 On average, our model's predictions are off by about  **$\pm 5.17$  minutes** compared to actual delays.

That means this flight could actually arrive anywhere from  **$\sim -5$  to  $\sim 22$  minutes delayed**, depending on real-world factors.


We used deep learning model trained on real flight data to provide the most accurate estimate possible.

# Gradio Workflow

## Sentiment Analysis



### Best Comment

 Trip Verified | Arrived at airport to find a 2 hour delay which is going to make us miss our connecting flight from Chicago house sitter. Guess no picnics for memorial day.

### Worst Comment

Not Verified | Rebooked my flight from Lubbock to Dallas stranding me in Dallas overnight. Fargo has very few flights from l



# Real-time Prediction



Predicts the flight arrival delay with a **95%** confidence interval and expected error.

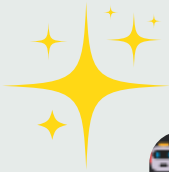
Confidence intervals provide clear expectations for delays.



# Real-time Prediction



## AI-based Recommendations



### AI-Generated Recommendation

1. Decrease the Departure Delay: Departure delay contributes significantly to the overall delay. In this case, it is 15 minutes aircraft.
  2. Improving Taxi-Out Time: The taxi-out time, which is the time from gate departure to take-off, is 10 minutes. This duration
  3. Realistic Scheduling: The scheduled elapsed time should be realistic and account for any possible contingencies. If the s
- Overall, these changes require an integrated effort involving ground staff, flight crews, maintenance teams, as well as better





**Thank you!**

