# **SQL 100 Days Challenge – Day 52 Reflection**

**Topic:** Airline Booking & Passenger Analytics

Dataset: Passengers, Flights, Bookings

### **Practice Experience:**

- Today's set of queries felt comparatively easy and smooth.
- Questions 1–10 mainly focused on revenue, cancellations, loyalty tiers, monthly trends, frequent flyers, and distance insights.
- I was able to use **LAG()** and **RANK()** functions efficiently without much difficulty these are concepts that earlier felt tricky but now come naturally.
- The **Bonus Challenge (Frequent Flyer Upgrade Eligibility)** was slightly advanced but still manageable, involving **CTEs**, **filtering with dates**, **and conditional logic**.

### **Key Learnings:**

- 1. **Revenue by Route:** Aggregated confirmed booking revenue per origin-destination pair.
- 2. **Top Passengers by Spend:** Ranked customers by spend to identify high-value travelers.
- 3. Cancellation Rates: Used CASE WHEN to calculate percentage of cancelled bookings.
- 4. **Ticket Pricing by Loyalty Tier:** Compared Gold, Silver, and Bronze loyalty members.
- 5. Monthly Revenue Trends: Applied LAG() to calculate month-over-month changes.
- 6. **Frequent Flyers:** Identified passengers booking more than 2 flights.
- 7. **Route Ranking:** Ranked flight routes using RANK() based on confirmed bookings.
- 8. **Booking Timeline:** Leveraged LAG() to track each passenger's booking history.
- 9. **Distance Insights:** Calculated miles flown per passenger with averages.
- 10. Country-Wise Contribution: Ranked countries by revenue contribution.
- 11. **Bonus Challenge:** Designed an upgrade eligibility rule (≥ 2 flights in 90 days + avg price > \$1000).

#### **Insights:**

- Indian passengers appeared prominently in revenue contribution.
- Gold-tier passengers often had higher average ticket prices.
- Cancellation rates varied by route, showing risk in long-distance flights.
- Frequent flyers with consistent high spend could be targeted for loyalty upgrades.

#### **Skills Reinforced:**

- Ranking & Window Functions (RANK, LAG)
- Aggregations with CASE WHEN
- Date filtering for last 90 days
- CTE usage for upgrade eligibility
- Business metrics: cancellations, revenue per route, loyalty analysis

### **Personal Note:**

Today felt **rewarding yet comfortable** — I could solve most questions quickly. It's encouraging to see how much smoother it is now to handle functions like LAG and RANK compared to earlier days. The bonus challenge was a good reminder of how SQL logic supports **real-world loyalty program strategies** in aviation.

## **Next Steps:**

- Explore customer segmentation by spend + loyalty tier.
- Expand monthly trends into year-over-year growth analysis.
- Create churn risk models for passengers with low booking frequency.