

# **FLT: SMARTER FUEL UPLIFT FORECASTING**

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# THE PROBLEM

F+ forecasts fuel uplift by averaging the **last 3 months** of actual uplift per route.

- Ignores schedule changes (cancellations, added flights)
- Can't predict new or seasonal routes
- Gets worse the further into the future it forecasts

Can we do better?

# VALIDATION SETUP

<b>Departure Airports</b>	BLL, FRA, VIE, PMI, ORD, HRG, KEF, HAM, HKG, WAW
<b>Airlines</b>	LH, OS, LX, SN, EW, EN, WK, 4Y, YF, XQ, 3S
<b>Routes</b>	~64 active
<b>Forecast Horizon</b>	6 months (Jul-Dec 2025)
<b>Total Volume</b>	1,610M kg fuel

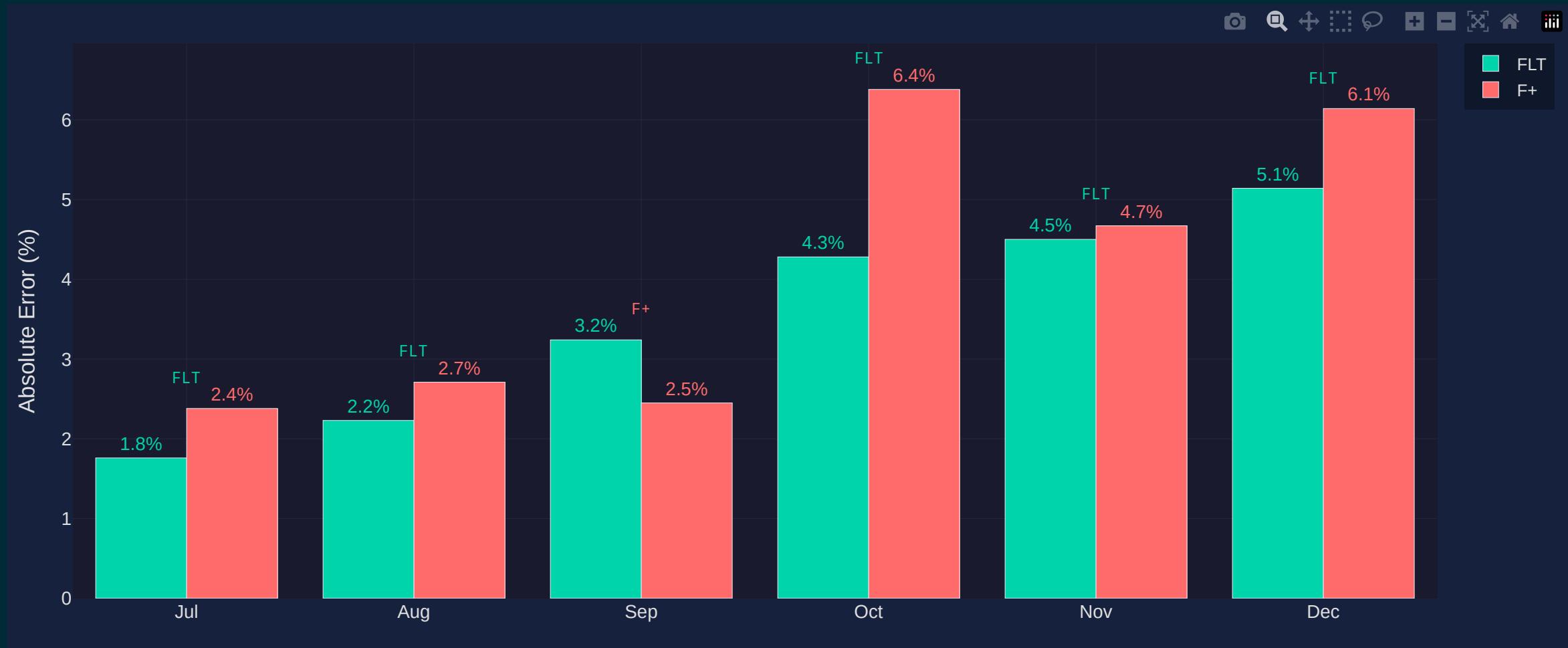
# FLT BEATS F+ IN 5 OUT OF 6 MONTHS

5 – 1

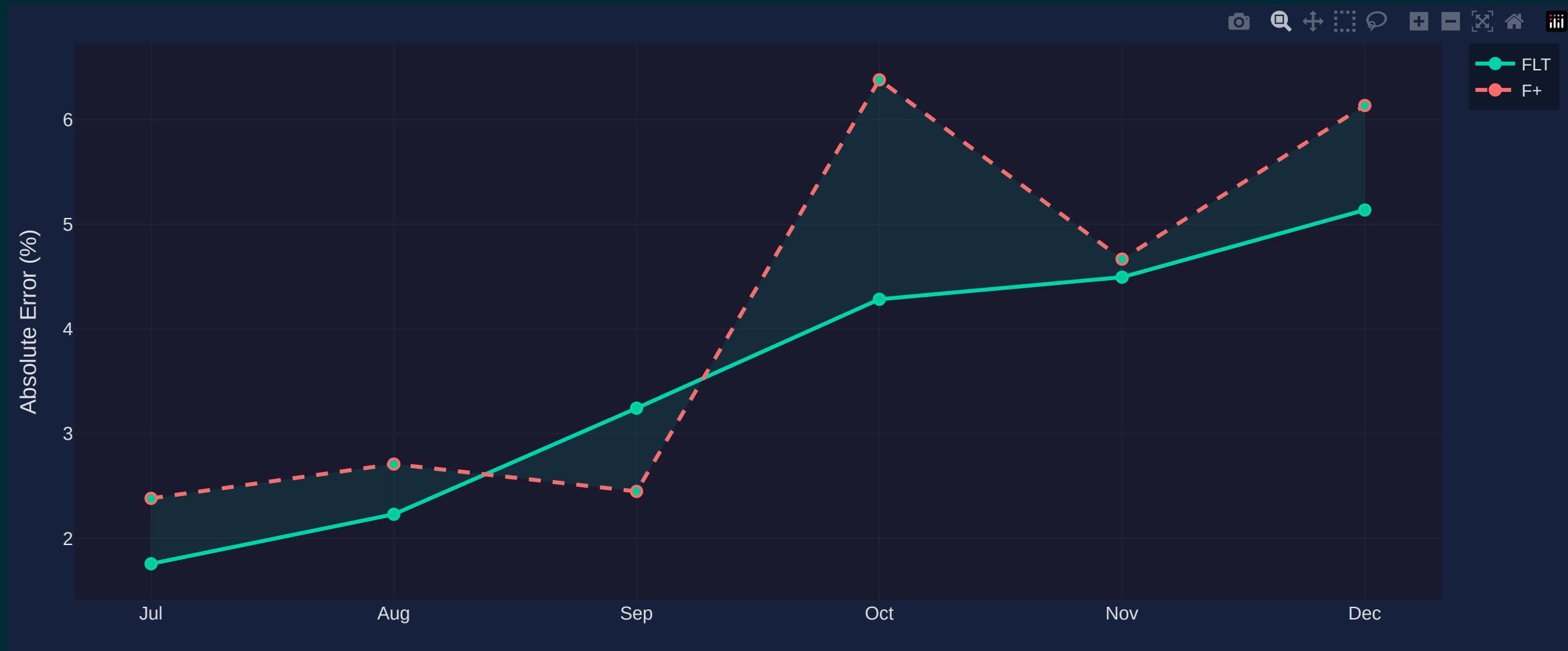


	FLT	F+
Overall Error	3.3%	3.9%
6-Month Misorder	53,800t	63,500t

# MONTH-BY-MONTH HEAD-TO-HEAD



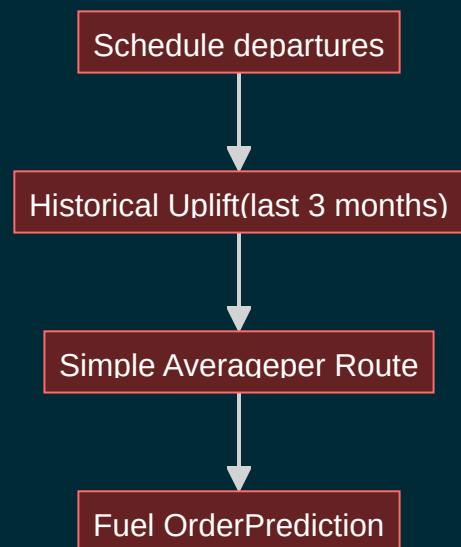
# THE FURTHER OUT WE FORECAST, THE BIGGER OUR EDGE



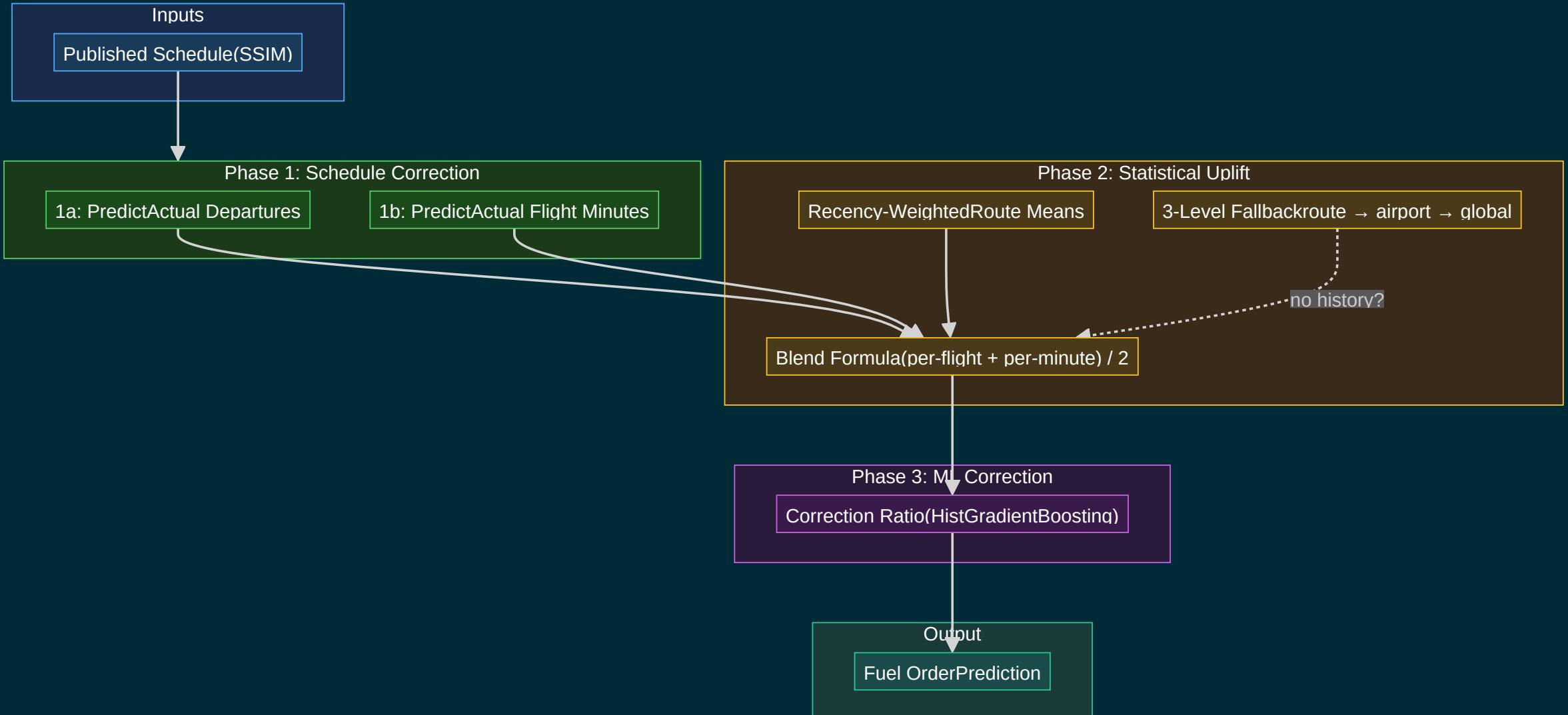
# WHY F+ STRUGGLES AT LONG HORIZONS

- F+ depends on the schedule and the previous 3 months uplift average
- When the schedule changes, F+ doesn't know
- **FLT corrects the schedule first**, then predicts fuel using ML
- Seasonal shifts, cancellations, and new routes are captured

# THE F+ PIPELINE



# THE FLT PIPELINE

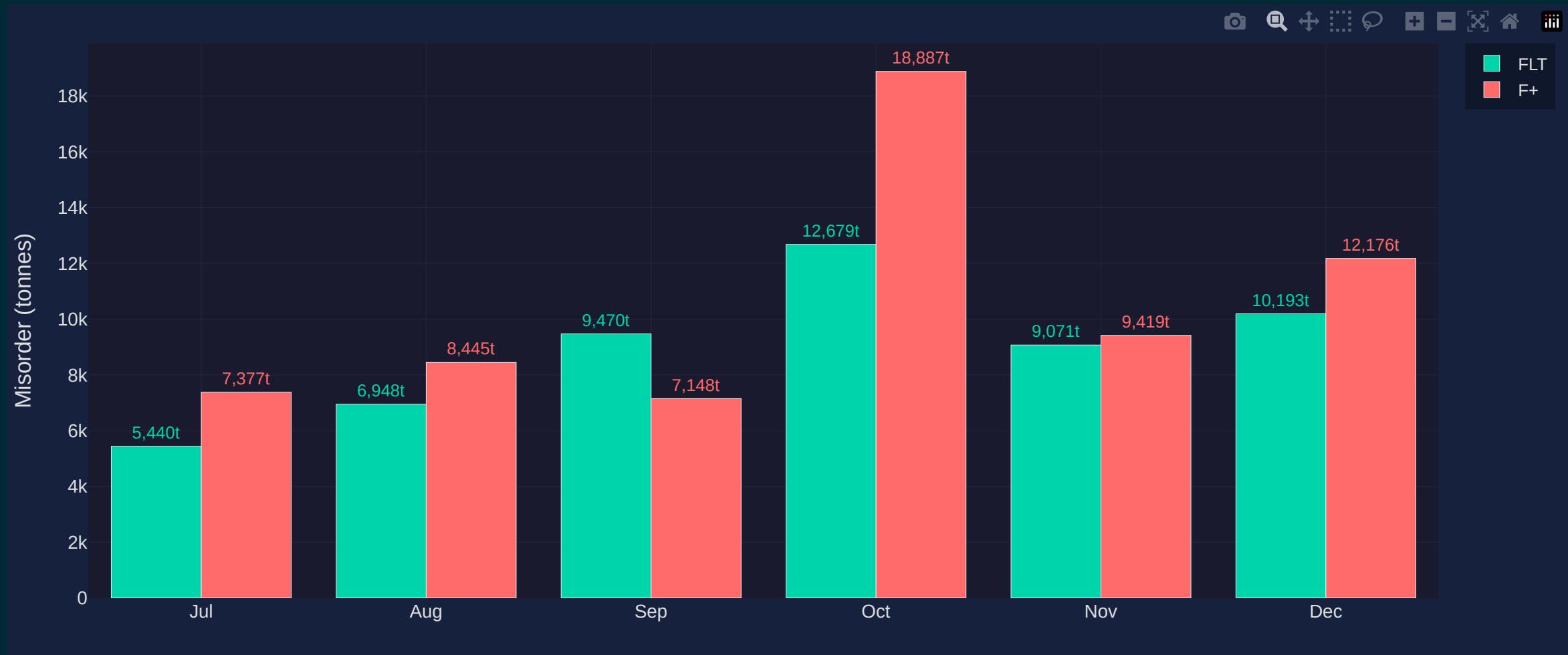


# HOW FLT WORKS

Step	What	Why
1a	Predict actual departures	Schedule is never perfectly right
1b	Predict actual flight minutes	Captures aircraft swaps & frequency changes
2	Statistical uplift formula	Recency-weighted route means x predicted volume
3	ML correction	Fine-tunes the formula at far horizons

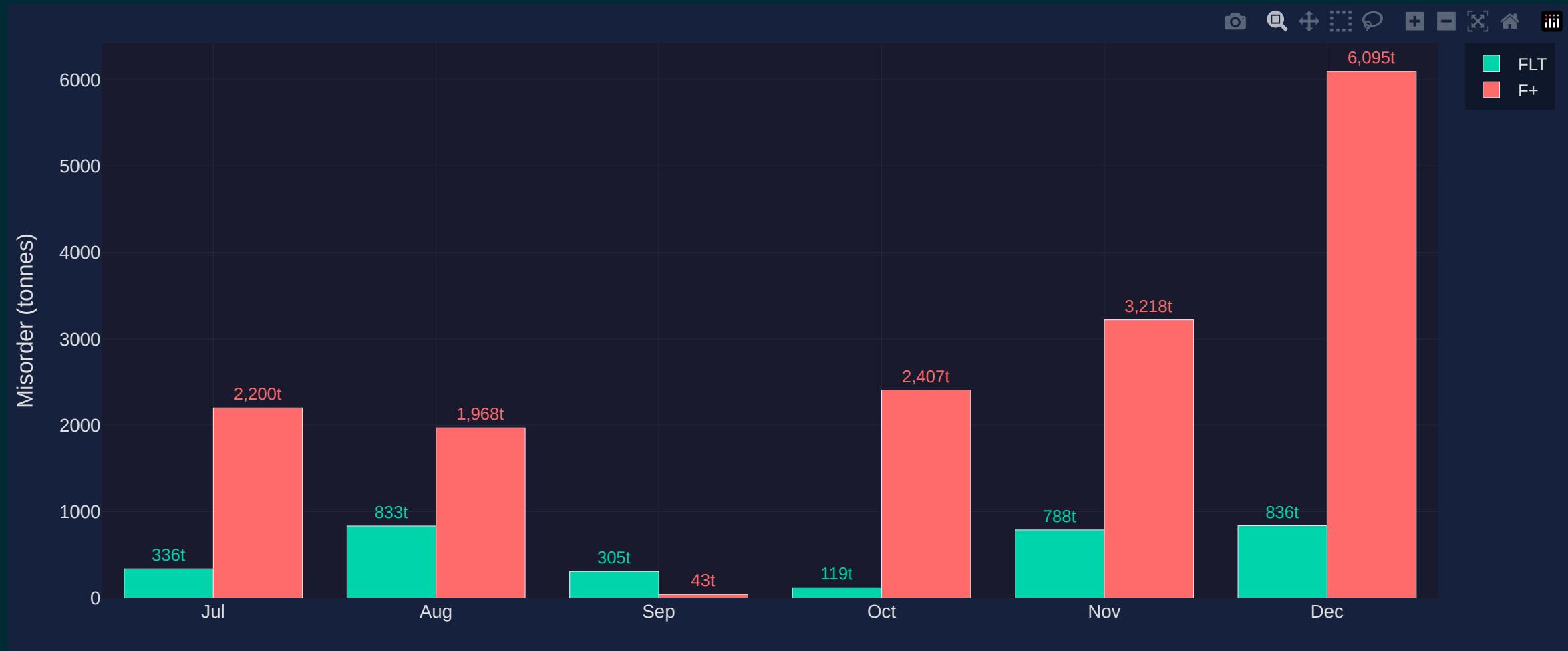
**Key insight:** We don't trust the schedule. We predict what will actually fly.

# FUEL ORDERING IMPACT



~9,700 tonnes less misorder than F+ over 6 months across 10 airports

# WHERE WE WIN BIG – LH-FRA



Even small % improvements on high-volume routes = large kg savings

# WHERE WE WIN BIG – UNSCHEDULED FLIGHTS

XQ-KEF, July 2025

	Scheduled	Actual	Predicted
Flights	0	2	2
Uplift (kg)	–	15,854	18,915

F+ has no prediction – it can't see what wasn't scheduled.

FLT caught the unscheduled flights using historical patterns.

# CURRENT SCOPE

Airports	BLL, FRA, VIE, PMI, ORD, HRG, KEF, HAM, HKG, WAW
Airlines	11 across Lufthansa Group
Horizon	6 months (Jul-Dec 2025 validated)
Routes	~64 active
Total Volume	1,610M kg fuel over validation period

This is our proof of concept. The model is ready to scale.

# ROADMAP

Next Steps	What	Deliverable
1	Multi-scenario validation	Models validated across diverse periods
2	Full airport expansion + ML pipeline	Complete coverage, automated training
3	Short-term production deployment	4-6 month predictions live, parallel with F+
4	Long-term model development	15-18 month predictions validated
5	Full system launch	All airports, all horizons, user interface

# THE OPPORTUNITY

- **Today:** 10 departure airports → **9,700 tonnes less misorder in 6 months**
- Every airport added means more routes, more data, more savings
- The model is airline-agnostic and built to scale
- ~400 airports on the roadmap

**More airports. Same model. More impact.**

## BUILT TO IMPROVE

- Trained on just **17 months** of schedule data (Jan 2024 – Jun 2025)
- The model has only seen **one summer, one winter season**
- With each additional month, patterns become stronger
- More airports = stronger shared patterns, without affecting existing route accuracy
- More data = better seasonal correction = fewer misorders

**These results are our floor, not our ceiling.**

# THE ASK

- **5 out of 6 months, 15% less fuel misorder**
- 10 airports, 11 airlines, 64 routes – 6 months validated
- Airline-agnostic: adding airlines = adding data
- Continue the project and expand coverage