Predicting Shipping Transit Times

Using Machine Learning & Gradient Boosting

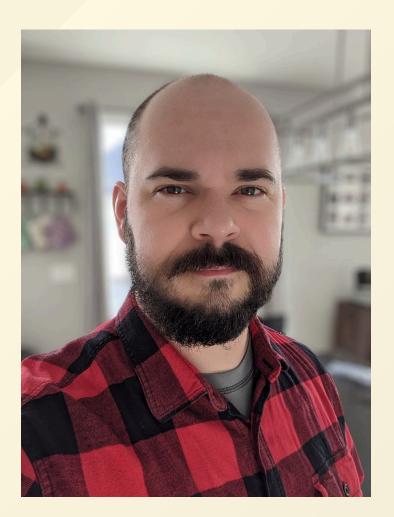
Our shipping data is more valuable than we realize...

Jakob Anderson

I've been building Machine Learning & other AI software for over 5 years, other software for 30 years.

AI & ML Apps that:

- Write stories
- Compose music
- Remix image styles
- Invent phrases
- Detect plagiarism



The Problem

Traditional shipping estimates fail:

- Based on simple lookup tables
- Inaccurate (±2-3 days variance)
- Lead to customer dissatisfaction
- Can't handle real-world complexity like weekends, holidays, days of the week, seasonal shifts

Business Impact:

- Lost revenue from missed deliveries
- Higher customer support costs
- Poor operational planning

Our Solution

73% more accurate predictions using ML

Metric	Traditional	ML-Based	Improvement
Accuracy	±2.5 days	±0.68 days	73% better
Speed	Slow	<50ms (cached)	Real-time

How it works:

Historical Data → Feature Engineering → LightGBM → Predictions

Powered by: 29,000+ shipments, 15+ features, gradient boosting

How Gradient Boosting Works

Builds trees sequentially, each correcting previous errors

Why LightGBM?

- Handles complex patterns automatically
- Fast: 10-50ms per prediction (cached)
- Robust to missing data
- Production-ready

Key Features Used:

- Temporal patterns (day/month)
- Routes & carrier combinations
- Seasonal historical trends
- Package characteristics

Increases Productivity

Before: Trusting the inaccurate carrier estimates, using carrier-provided charts that don't accommodate for seasons, weekends, zones

After: Data-driven predictions for better carrier comparison

Benefits:

- Additional metric for rate shopping decisions
- Compare predicted transit times across carriers
- Make informed tradeoffs between cost and speed
- Batch processing for large shipment volumes

├─ Improves Work Quality

Precision & Reliability:

- 73% more accurate predictions
- Uncertainty ranges (best/expected/worst case)
- Data-driven decision making
- Statistical distribution analysis

Results:

- Fewer missed delivery windows
- More accurate customer promises
- Better resource allocation

© Unlocks New Capabilities

Uncertainty Quantification:

• Best case: 2.1 days

• Expected: 3.2 days

• Worst case: 4.8 days

New Services:

- Offer guaranteed delivery windows
- Real-time analytics dashboard
- Carrier performance comparison
- Paid API for third-party apps

Makes Money

Speculative Business Value of Accurate Predictions:

- Paid service for third-party apps. Use predictions from our shipping data as a valuable resource to inform other's decisions.
- Avoid costly late deliveries and lost sales.
- Prevent loss of customer trust and reduce the need for service credits or refunds.
- Less wasted resources, fewer emergency shipments, and improved inventory management.
- On-time shipments can be a key differentiator in competitive markets.
- Loss aversion: Businesses are more sensitive to losses from missed deadlines than to gains from faster delivery
 - -ML helps minimize these risks.

Bottom Line:

• Accurate transit time prediction is a strategic asset that protects revenue, reduces operational losses, and enables new business opportunities.

Beyond Transit Time: Future Opportunities

Logistics:

- Demand forecasting for resource planning
- If we know which carriers/levels slow down for each season, we can route shipments to those that don't slow down
- Dynamic pricing based on capacity

Business Intelligence Uses for Data Mining & ML predictions:

- Customer churn prediction
- Fraud detection with real-time scoring
- Risk assessments
- Many other paid api prediction services using historical shipping data

Next Steps

Current System:

- ✓ Trained models with 73% better accuracy than carrier charts
- ✓ REST API
- Analytics dashboard deployed

Your Pilot Project:

- 1. Integrate with your shipment data
- 2. live proof of concept, showing informative estimate
- 3. Measure ROI on real shipments
- 4. Scale across operations

Questions?