

# Adaptive Route Optimization with InTandem

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## Problem Statement & Real-World Use Case

Shipping routes today are mostly static, even when real-world disruptions (e.g., traffic, storms, customs delays) arise; this leads to delayed or costly deliveries. **InTandem** simulates a smarter algorithm that adjusts in real time, re-routing based on live disruption data to optimize delivery time, cost, and risk.

## Test Plan & Results

### **Test 1** – *No Disruptions*

Graph:  $A \rightarrow B \rightarrow C$  (cost 5),  $A \rightarrow C$  (cost 12)

→ Expected: 5

→ Actual: 5

### **Test 2** – *Moderate Traffic Disruption on $B \rightarrow C$*

Disruptions: {10, 4, 0, 0}

→ Expected: ~9

→ Actual: 9

### **Test 3** – *Extreme Disruption Avoidance*

$B \rightarrow C$  highly penalized,  $A \rightarrow D \rightarrow C$  safer

→ Expected: 6

→ Actual: 6

### **Test 4** – *Live-Randomized Inputs*

Disruptions generated at runtime

→ Output varies dynamically with penalties applied.

**Output:**

```

-----
Test 1: Base Shortest Path without Disruptions
Expected shortest cost to C: 5 (A->B->C: 2+3)
Actual shortest cost to C: 5
Correctly routes A->B->C (5) instead of A->C(12)
-----

Test 2: Weighted Disruption Penalty Influences Routing
Expected shortest cost to C: 9 (2 + adjusted ~7)
Actual shortest cost to C: 9
Correctly routed A->B->C (~9) instead of A->C (12)
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Test 3: Extreme Disruption Only Affects B->C
Expected shortest cost to C: 6 (A->D->C)
Actual shortest cost to C: 6
Correctly routed around high-penalty B->C path via A -> D -> C
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Test 4: Live Randomized Disruptions at Node C
Generated disruptions for B -> C:
Traffic: 3
Weather: 10
Fuel: 5
Customs: 4
Shortest cost to C: 5

```

## Core Algorithm

```

for (auto& edge : graph[node]) {
    int adjusted = edge.baseWeight;

    // calculate weighted disruption score for this specific edge
    if (!edge.disruptions.empty()) {
        double weightedScore = 0.0;
        for (size_t i = 0; i < weights.size() && i < edge.disruptions.size(); ++i) {
            weightedScore += weights[i] * edge.disruptions[i];
        }
        if (weightedScore > 5) {
            // if the combined disruption score exceeds the threshold (5),
            // apply a quadratic penalty to the base cost to reflect severity
            adjusted = static_cast<int>(round(edge.baseWeight * (1 +
pow(weightedScore - 5, 2))));
        }
    }

    // update shortest path if this new adjusted cost is better
    if (cost + adjusted < dist[edge.to]) {

```

```
        dist[edge.to] = cost + adjusted;
        pq.push({dist[edge.to], edge.to});
    }
}
```

## Runtime & Memory

Modified Dijkstra's algorithm.

### Time Complexity:

- $O((V + E) \log V)$  using a min-heap priority queue, where:
- $V$  = number of nodes (locations)
- $E$  = number of edges (shipping routes)
- $O(1)$  - assuming a fixed number of disruption categories

### Space Complexity:

- $O(V)$  for distance map `dist`
- $O(E)$  storing graph adjacency list
- $O(V)$  priority queue
- Total =  $O(V + E)$

## Discussion: Tradeoffs & Future Work

All disruption types are merged into a single score, which simplifies logic but loses specificity. It doesn't consider route-specific thresholds, per-shipment urgency, or multi-objective goals (fastest vs cheapest).

Future work includes:

- Time-window constraints
- Multi-shipment load balancing
- Scaling to 100+ node networks
- Real API hooks for live event data
- Route-level disruption customization

## Github Repo

<https://github.com/anhpls/intandem>

**Release Tag:** v1.0-final

## **Link to Presentation**

<https://docs.google.com/presentation/d/1q52GqUrGiFjkLEI7kY9xVXLBqCAOHKBombI4TnKsZoM/edit?usp=sharing>