





# TECHTRANSIT PROBLEM - II



Here is where your trip begins









## **Overview of the Problem**

This problem requires a system that can evaluate different multimodal transport routes and optimize for cost, time, and feasibility. Here's a structured approach:

#### **Solution Overview**

Develop a Multi-Modal Route Selector that takes shipment details as input and returns optimal routes based on:

- Modes of Transport: Air, Sea, Land, and combinations
- Cost Estimation: Freight charges, customs duties, fuel costs, handling fees
- Transit Time: Estimated delivery times for each mode
- Border Feasibility: Customs regulations, restrictions, and delays





# **Overview of the Problem**

#### **Key Components**

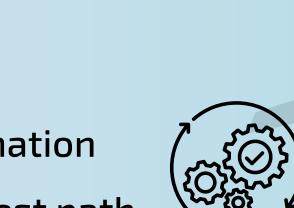
#### **User Input:**

- Origin & Destination (Country, Port, Airport, City)
- Cargo Details (Weight, Volume, Type)
- Preferred Priority (Cost, Speed, or Balanced)

### **Route Optimization Engine:**

- Uses a database of transportation routes
- Incorporates real-time shipping data for cost & time estimation
- Implements an algorithm (Dijkstra/A)\* to determine the best path
- Factors in border-crossing regulations & trade agreements











# **Overview of the Problem**

# **Output:**

- Ranked list of optimal routes (e.g., Air-Sea, Land-Air)
- Cost, Estimated Time, Border Checkpoints
- Visual Route Map



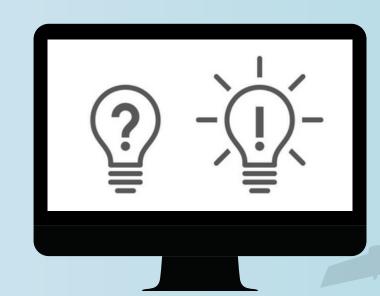






# **Current Solution**

- Objective: Analyze and compare various route optimization algorithms to improve route planning systems.
- Algorithms Covered:
  - Traditional: Dijkstra's, A\*, Genetic Algorithms.
  - Modern: Ant Colony Optimization, Particle Swarm Optimization, Machine Learning-based techniques.
- Evaluation Criteria:
  - Computing efficiency.
  - Scalability.
  - Adaptation to changing environments.
  - Accuracy in real-world scenarios.
- Methods Explored:
  - Heuristics.
  - Hybrid approaches.
  - Parallel computing.
- Applications: Practical use in urban planning, logistics, and transportation.
- Validation: Case examples and experimental data to validate theoretical conclusions.
- **Outcome**: Provide resources for practitioners, researchers, and decision-makers to choose the best route optimization method based on specific application needs.





# **Pitfalls of the Current Solution**



#### **Computational Complexity:**

Modern algorithms like Ant Colony
Optimization and Particle Swarm
Optimization can be computationally
intensive, requiring significant processing
power and time.

#### **Scalability Issues:**

Traditional algorithms such as Dijkstra's and A\* may struggle with scalability when applied to large-scale networks or realtime applications, limiting their effectiveness.

# Adaptation to Dynamic Environments:

Many algorithms do not adapt well to rapidly changing conditions, such as real-time traffic updates or sudden changes in logistics requirements, leading to suboptimal route choices.

#### **Integration Challenges:**

Integrating these algorithms into existing systems can be complex and may require significant modifications to current infrastructure, posing a barrier to implementation.

#### **Data Dependency:**

The effectiveness of machine learningbased techniques heavily depends on the quality and quantity of available data, which can be a limitation in some scenarios, affecting the accuracy of route optimization.



