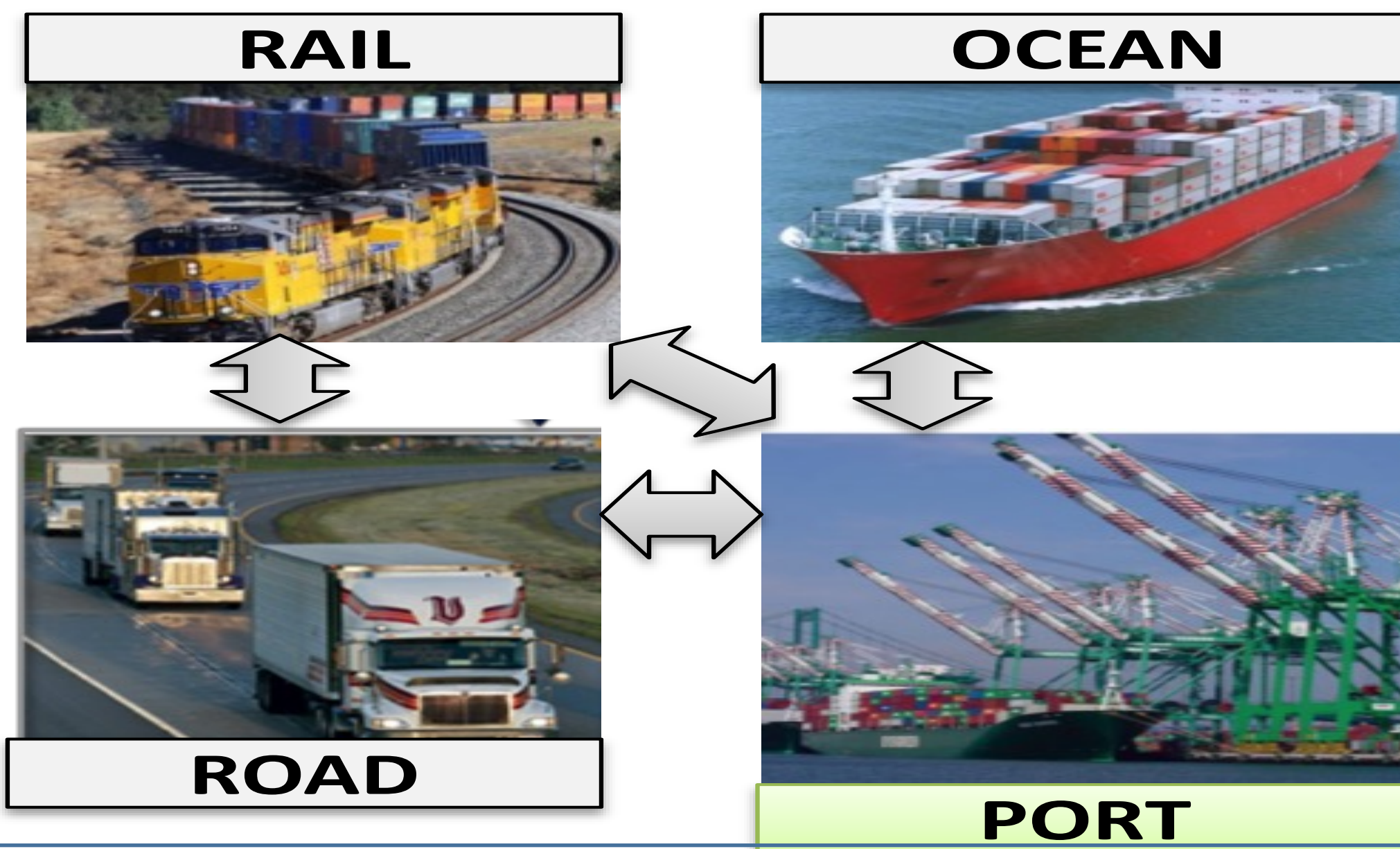
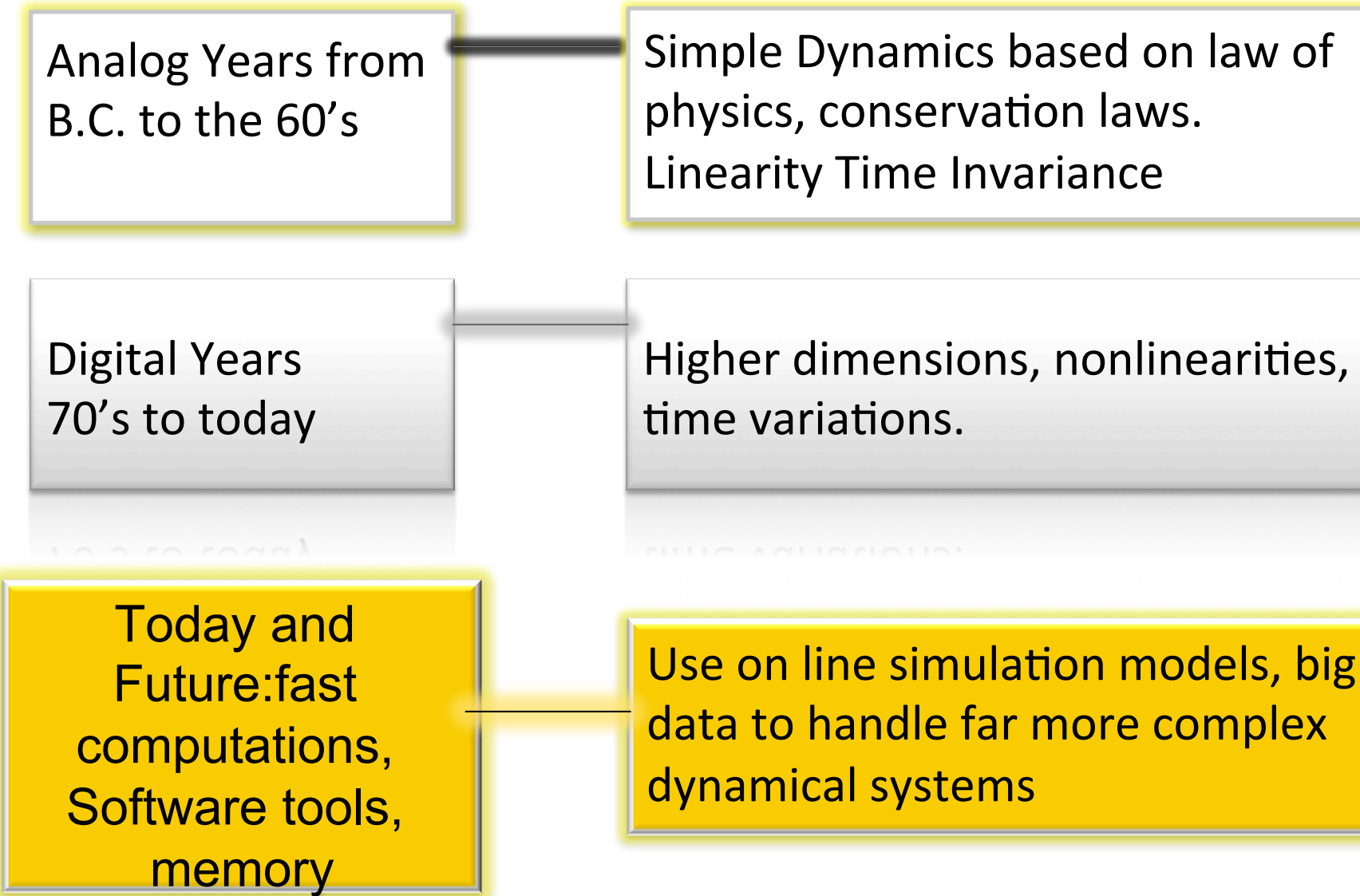


Motivation and Objective

Complex Freight Transportation System: How to model, optimize and control



Model Evolution

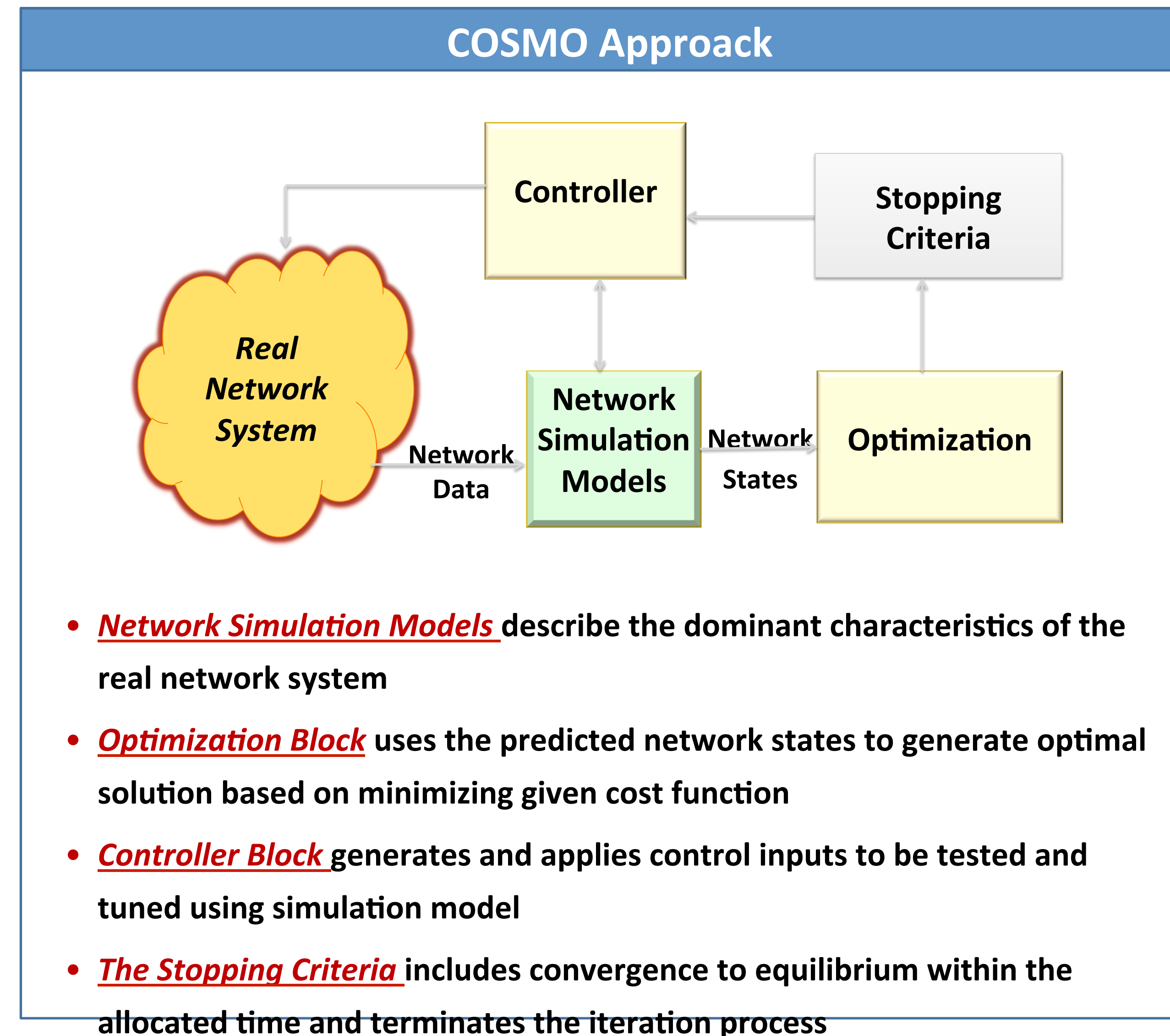


- Simulation models can be used in real time to predict the states of the network by fast forwarding based on current and expected demand

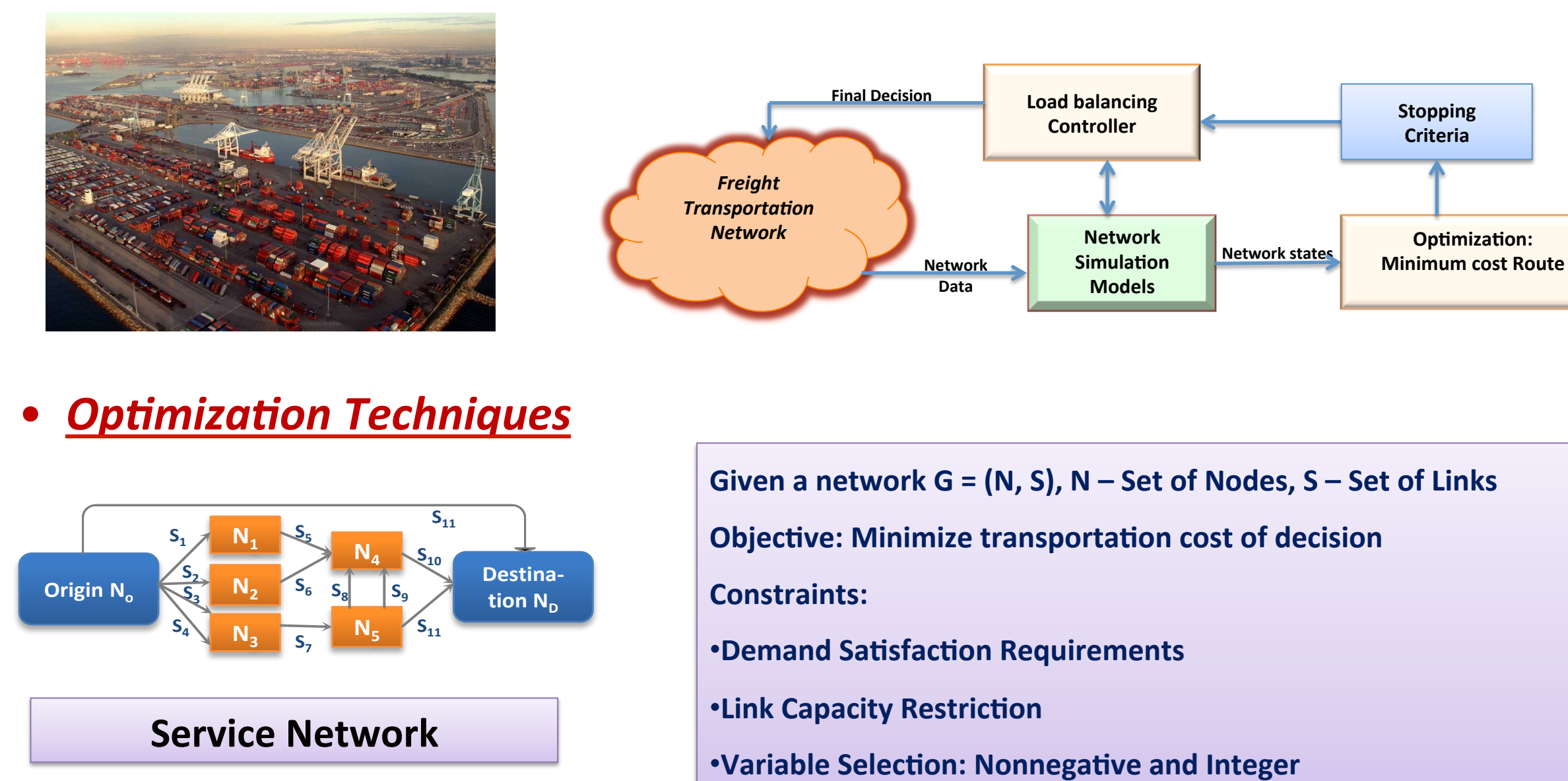
Project Objectives

- Develop the theoretical foundations of a new control approach referred to as **COSMO (Co-Simulation Optimization, Control)** to optimize and control complex dynamical networks with temporal and spacial characteristics.
- Investigate stability, convergence, robustness and scalability issues
- Use multimodal freight load balancing as the application area
- Investigate how identified barriers and policy issues/incentives can be incorporated as mathematical constraints and/or control variables in the optimized dynamic freight load balancing system
- Use real data to validate the simulation models and generate realistic scenarios for validation and testing using a testbed for the Los Angeles/Long Beach area that involves interacting road/rail/port networks
- Assess the effectiveness and feasibility of implementing freight load balancing strategies to increase the efficiency and sustainability of urban freight movements by interacting with participating stakeholders
- Integrate the research results to the University educational program by training students to new problems both in theory and application to freight transportation

Co-Simulation Optimization (COSMO) Approach



COSMO for Dynamic Freight Load Balancing



$$\min \sum_{t \in T} \sum_{(i,j) \in S} \sum_{p \in P} c_{ij}^p x_{ijt}^p + \sum_{t \in T} \sum_{(i,j) \in S} f_{ijt} y_{ijt}$$

Subject to:

$$\sum_{t \in T} \sum_{j \in N} x_{ijt}^p - \sum_{t \in T} \sum_{j \in N} x_{jlt}^p = d_{it}^p, \quad i \in N, \quad \forall p \in P$$

$$\sum_{p \in P} x_{ijt}^p \leq u_{ij} y_{ijt}, \quad (i,j) \in S, \quad \forall t \in T$$

$$x_{ijt}^p \geq 0, \quad (i,j) \in S, \quad \forall t \in T, \quad \forall p \in P$$

$$y_{ijt} \in \{0,1,2,\dots\}, \quad (i,j) \in S, \quad \forall t \in T$$

f_{ijt} is the cost per service unit incurred by using service (truck or train) on link (i,j) at time t ,
 y_{ijt} is the discrete decision variable which represents number of service units offered on link (i,j) at time $t \in \{1,2,\dots,T\}$. c_{ij}^p is the transportation cost per unit flow of product p on link (i,j) . x_{ijt}^p is the amount of flow of commodity p using link (i,j) , d_{it}^p is the demand of product p at node i , and u_{ij} is the capacity of link (i,j) .

Optimization Algorithm Candidates

- Decompositions + Branch and Bound Techniques
- Natural heuristics

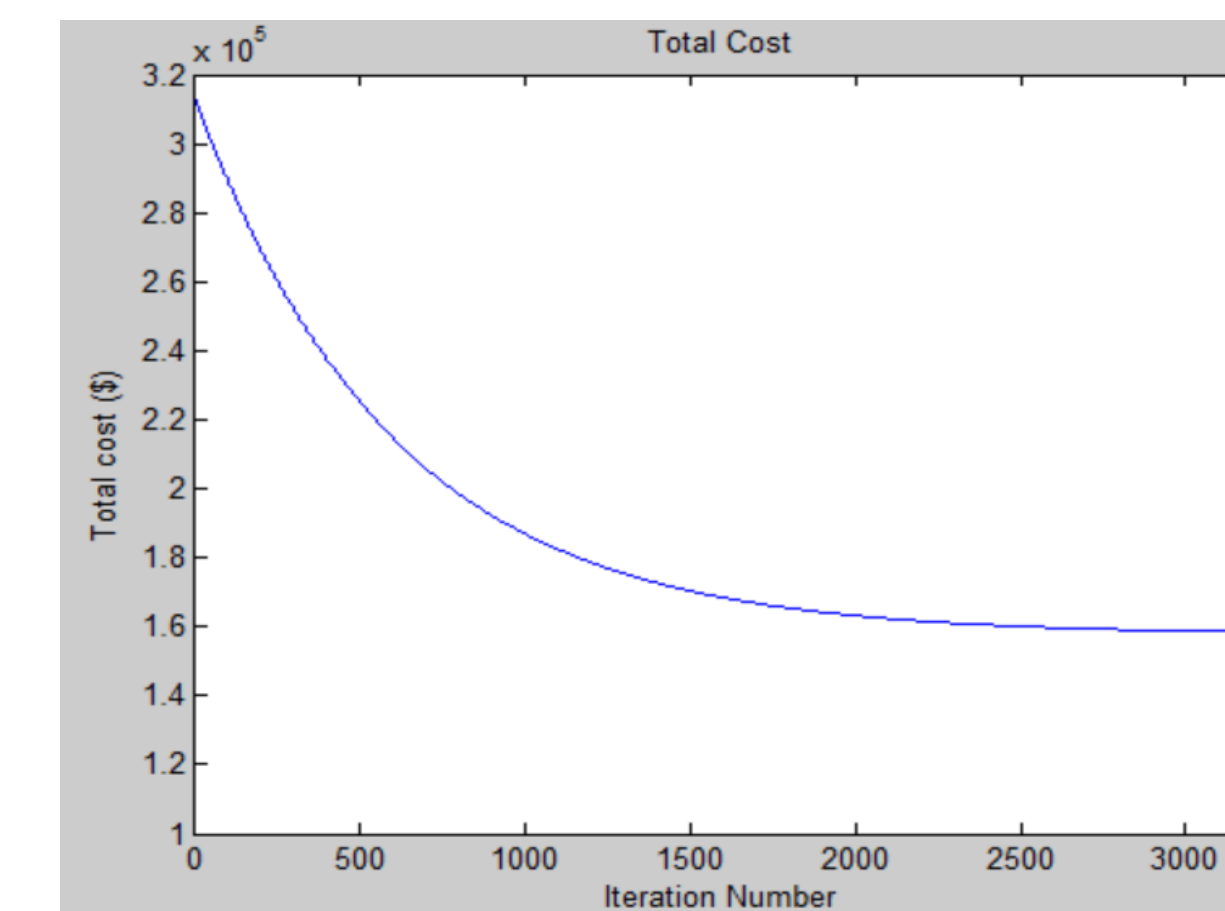
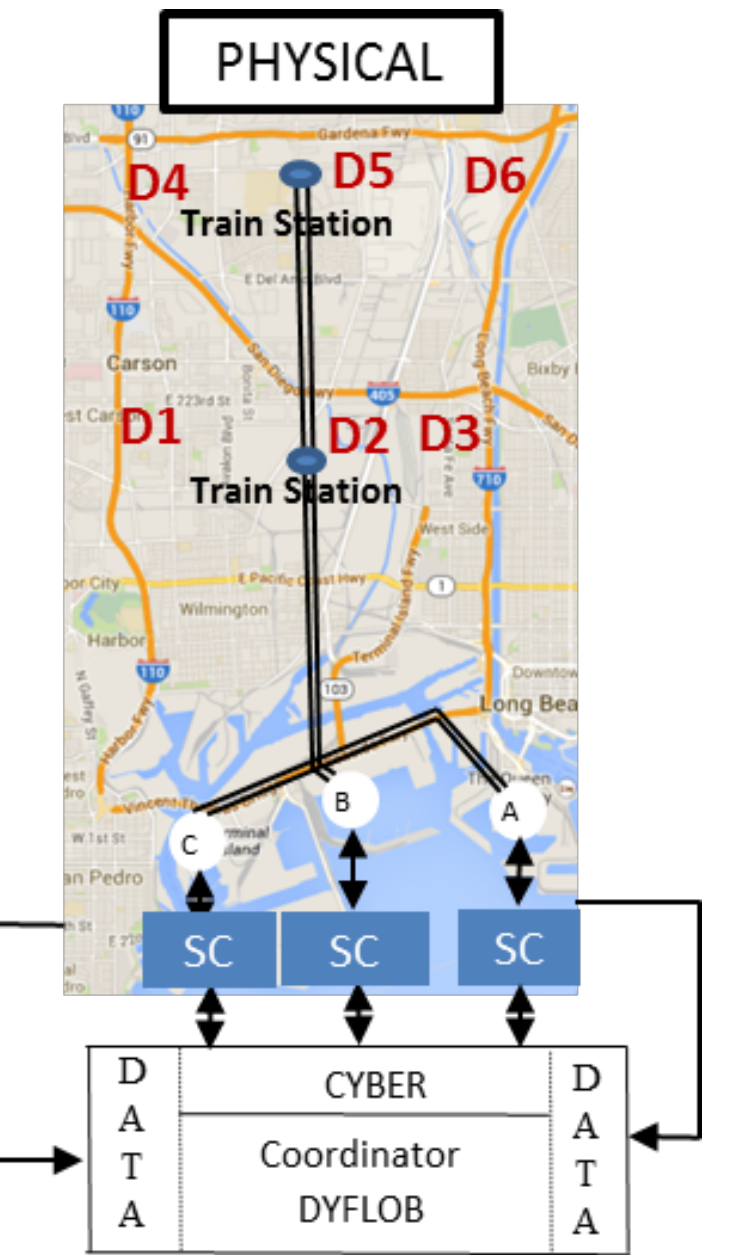
Preliminary Results: Load Balancing Example

Physical Network

About 80 square miles near the Port of LA/LB
 Road network with 12824 links and 4747 nodes plus rail
 Traffic and rail data from the Southern California Association of Governments (SCAG)

Evaluation Scenario

Origin: 3 terminals A (1020), B(1020), C(1020)
Destinations: D1 (350), D2(450), D3(400), D4(600),D5(700), D6(560)
Shippers: Three, one for each terminal
Routes: Road Network and Partial Rail (5 trains with capacity of 50 containers each)
Simulation Model: Macroscopic based on VISUM. Use historical data and dynamic assignment to tune it.
Cost: Minimize total \$ cost based on real data which depends on travel times and mode



Load balancing iterations that reduce the overall cost

The experiment was performed on a computer with the Intel 2.3 GHz CPU and a standard routine from Matlab is used to solve the optimization problem at each iteration.

The load balancing control procedure used is rather adhoc and involves shifting loads from previously minimum cost routes to the new minimum cost route generated by the optimizer till all routes have equal cost leading to new states of the network generated by the simulation model which are fed to the optimizer for a solution under the new situation.

Validation using Testbed

