CPS: Medium Collaborative Research: Smart Freight Transport Using Behavioral Incentives

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Motivation

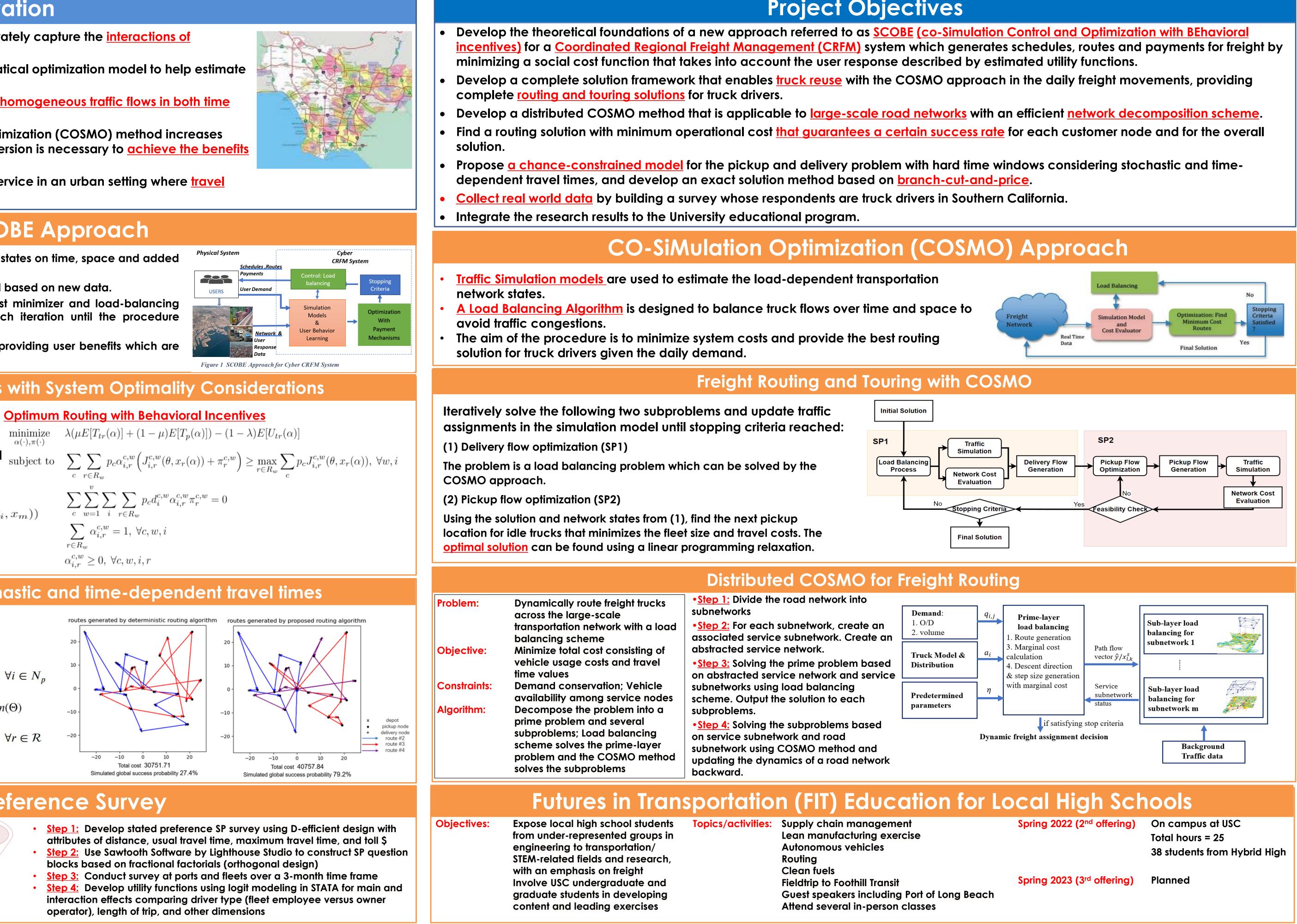
- Modeling based on pure mathematical formulations cannot accurately capture the interactions of passenger and truck flows. • Therefore, traffic simulation models are introduced into a mathematical optimization model to help estimate the transportation network states. • Better truck routing solutions can be generated by balancing non-homogeneous traffic flows in both time and space. • The computational complexity of a centralized CO-SiMulation Optimization (COSMO) method increases exponentially with the size of the targeted network. A distributed version is necessary to achieve the benefits of algorithmic performance and computational speed. • It is essential to improve the reliability of the pickup and delivery service in an urban setting where travel times are stochastic and time dependent. **Proposed SCOBE Approach** Simulation models are used to generate/predict the dependencies of link states on time, space and added loads. Utility functions for each user or class of users are established and updated based on new data. converges or a stopping criterion is satisfied. attractive enough to attract participation. Personalized Route Recommendations with System Optimality Considerations • Step 1: Cluster the drivers into distinct groups $\underset{\alpha(\cdot),\pi(\cdot)}{\text{minimize}}$ • <u>Step 2:</u> Learn the utility function of each cluster • <u>Step 3:</u> Solve the optimum routing problem that provides behavioral incentives <u>Utility learning</u> minimize $\mathcal{L}(\theta_i, x, y)$ where $\mathcal{L}(\theta_{i}, x, y) = -\frac{1}{M} \sum_{i=1}^{M} y_{m} log(s(\theta_{i}, x_{m})) + (1 - y_{m}) log(1 - s(\theta_{i}, x_{m}))$ $s(\theta_{i}, x_{m}) = \frac{1}{1 + \exp(J_{i}(\theta_{i}, x_{m1}) - J_{i}(\theta_{i}, x_{m2}))}$ Pickup and delivery problem with stochastic and time-dependent travel times The model was formulated as a chance-constrained integer $minimize \quad \sum_{r \in \mathcal{R}} c_r x_r$ optimization problem with a setpartitioning formulation. s.t $\sum_{r \in \mathcal{R}} a_{ir} x_r = 1$ $\forall i \in N_p$ An exact algorithm based on a <u>branch-</u> <u>cut-and-price</u> was proposed to solve $\sum_{r \in \mathcal{R}} x_r \ln(P^r) \ge \ln(\Theta)$ the problem. In the pricing problem, **a** new labeling algorithm and dominance $x_r \in \{0, 1\}$ $\forall r \in \mathcal{R}$ rules are designed to deal with stochastic travel times and probabilistic information. **Driver Stated Preference Survey Objectives:** Obtain data from short-haul drayage truck drivers on behaviors and preferences in choosing delivery routes Develop incentives for alternative trip route/time of day from estimated utility
- <u>Utility function information</u> and <u>payment incentives</u> are used by the cost minimizer and load-balancing controller in an iterative procedure that leads to a lower cost at each iteration until the procedure
- The aim of the central coordinator is to achieve system optimality while providing user benefits which are



functions and value of time

<u>Route choice</u>

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blocks based on fractional factorials (orthogonal design) **<u>Step 3:</u>** Conduct survey at ports and fleets over a 3-month time frame <u>Step 4:</u> Develop utility functions using logit modeling in STATA for main and interaction effects comparing driver type (fleet employee versus owner

Project Objectives

dents ips in i/ arch,	Topics/activities:	Supply chain management Lean manufacturing exercise Autonomous vehicles Routing Clean fuels	Spring 2022 (2 nd offering)	On campus o Total hours = 2 38 students fro
and bing es		Fieldtrip to Foothill Transit Guest speakers including Port of Long Beach Attend several in-person classes	Spring 2023 (3 rd offering)	Planned

