

Arcology Simulation Framework

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Master of Science Thesis

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Project Summary:

Optimization and simulation framework to analyze transit-oriented designs

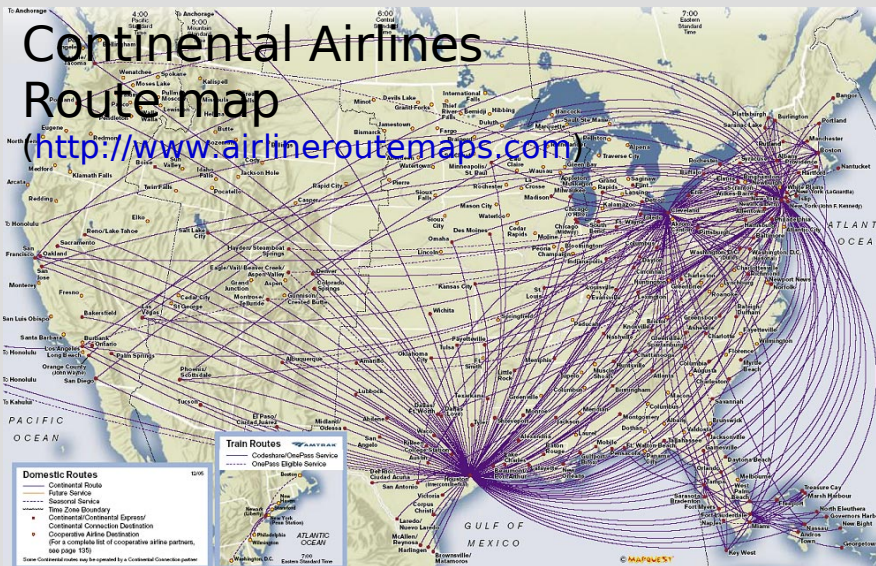
Address 2 questions:

1. How can we evaluate the effectiveness of an urban complex?

- Demand / Sustainment / Measurement framework:
 - Investigates demand distribution patterns influenced by urban planning topology
 - Quantifies effects of transportation infrastructure topology and mode of operation
 - Determines system's ability to satisfy resident / industrial needs

2. What transit paradigms succeed at making the world “smaller”?

Mass Transit Paradigms: Commercial Aviation



- Hub-and-Spoke
 - economies of scale with mixed fleets
 - 767 & 757
- Point-to-Point
 - more direct flights with fleets of regional jets
 - SWA 737
- SATS
 - service from small local airports could take Point-to-Point concept to an extreme



Ground Transit establishes Feeder-and-Trunk model

- Bus routes often feed subway / light rail trunks
 - connecting to other modes of transportation
- HCPPT shows the capability of a more distributed demand-responsive model

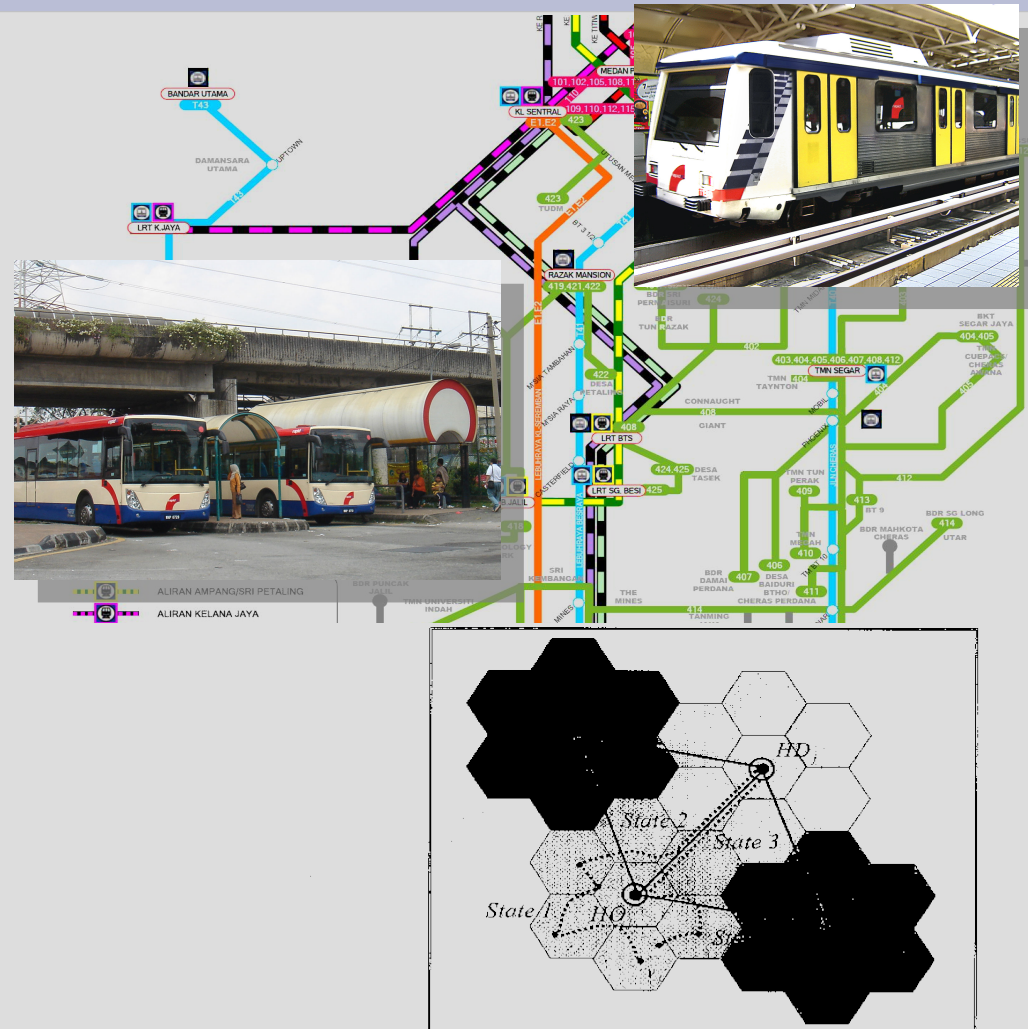
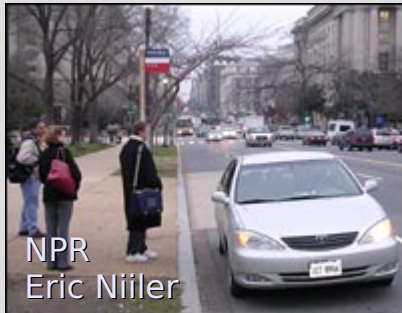


Figure 4.9 Vehicle schedule representation

(Cortes 2003 *HCPPT: A New Design Concept and Simulation-Evaluation of Operational Schemes*)

Vehicle Sharing Options and Concepts



- Carpools / HOV Slugs
- Flexcar / Zipcar rental services
- Taxi cab network
- Robotic driverless cars
- CityBike Amsterdam GPS bicycle system

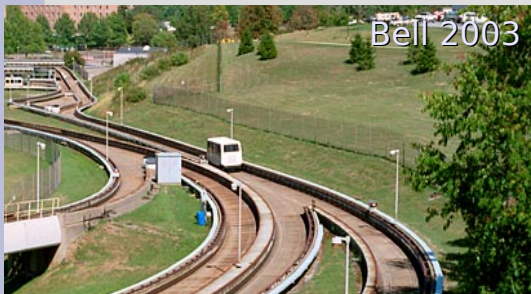
Personal Rapid Transit Systems struggle along



CabinTaxi verified and tested in Germany, abruptly abandoned due to NATO commitments



Taxi2000 branched from Raytheon



Morgantown, WVU operational group transit system; abandoned by Boeing



ULTra system slated for 2007 deployment in Heathrow airport, UK and Dubai, UAE

7 Transit Oriented Design should drive development of more efficient mass transit

- We often search for advanced transportation solutions to energy problems
 - We can make larger impacts by reducing travel need/distance by adjusting urban planning and logistics
- Urban Layout
 - Increase density
 - Culminating in arcology concepts
 - Increased density correlated with decreased energy use per capita
- Logistics
 - Stagger work schedules to reduce peak loads
 - Flexibility to optimize residence / workplace pairings
 - Mass transit effectiveness that rivals personally-owned vehicles in door-to-door performance
 - Enabled by transit-oriented design



Denser cities are more efficient per capita

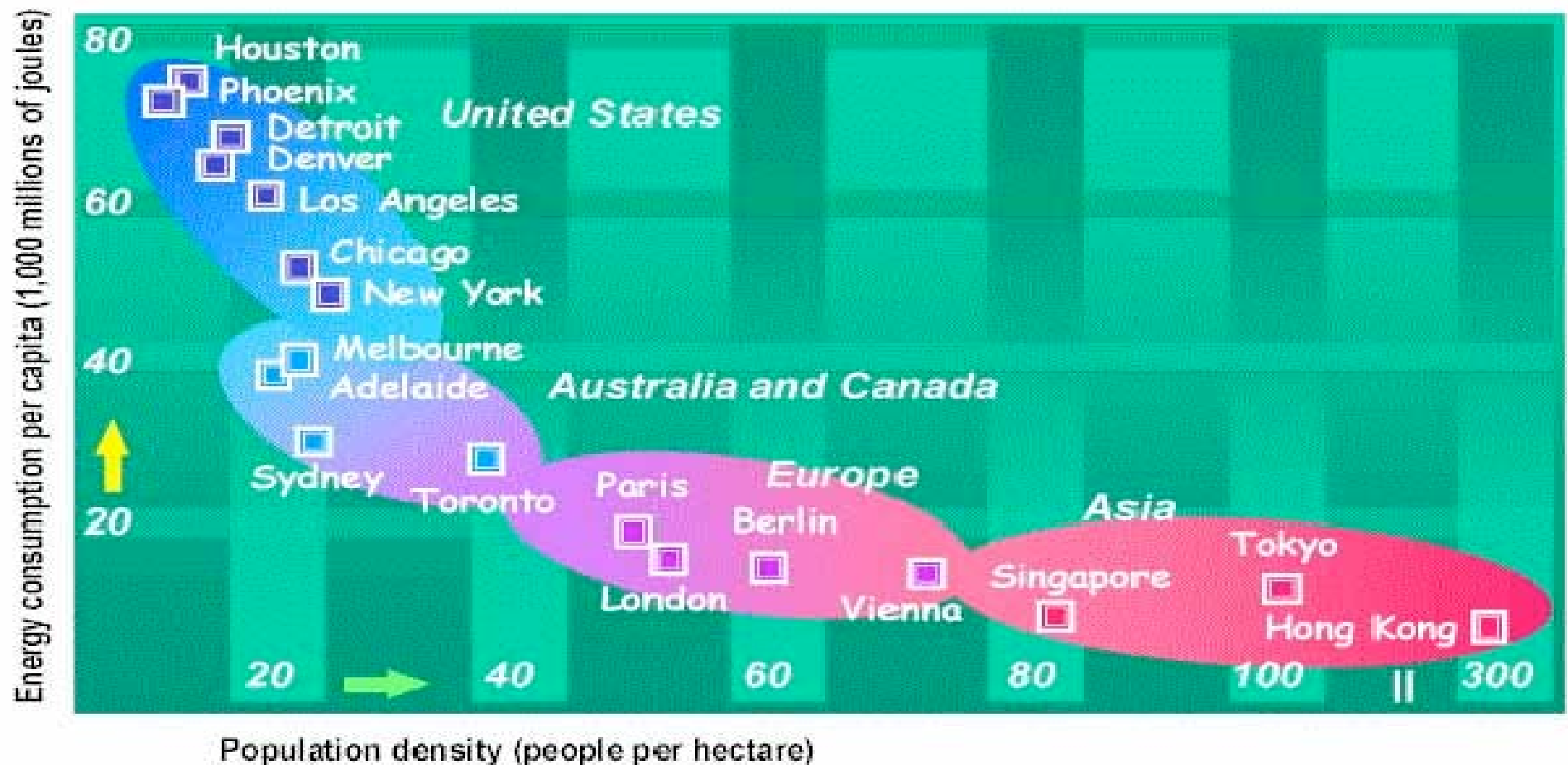


Figure 2. The relationship between population density and energy consumption in cities.

(Emmi 2003 *Coupled Human–Biologic Systems in Urban Areas: Towards an Analytical Framework Using Dynamic Simulation*)

Arcologies and Compact Cities pack functionality



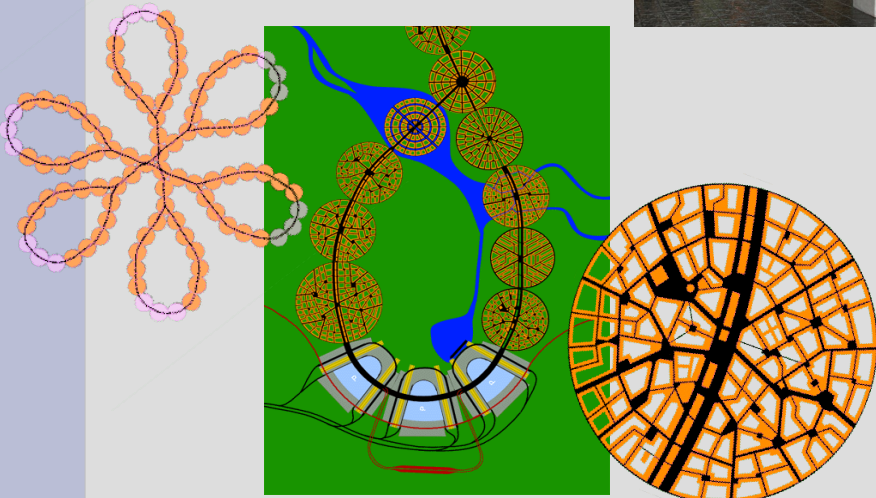
Soleri's *Arcology*

- Architectural implosion of cities
- Form a human relationship to the environment

Dantzig & Saaty's *Compact City*

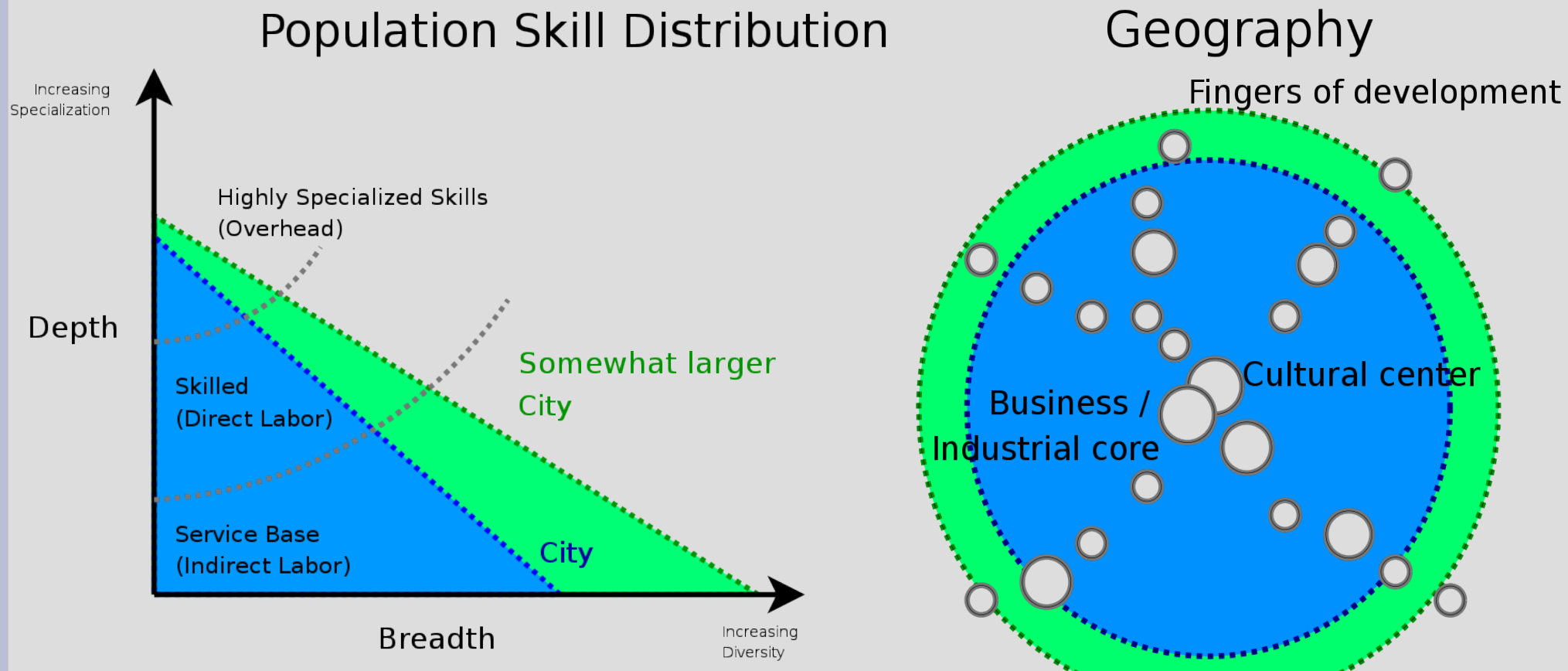
- Comprehensive proposal for many aspects of a functioning hyperstructure

- Crawford's *Carfree Cities*
 - Reference designs most applicable to transit approach and assumptions used in this thesis



A Metropolitan complex should maximize diversity

- Offer diverse set of specialized skills and jobs
- Well-suited for a systems approach to the design of life support infrastructure



Mass Transit Optimization

Key Capabilities

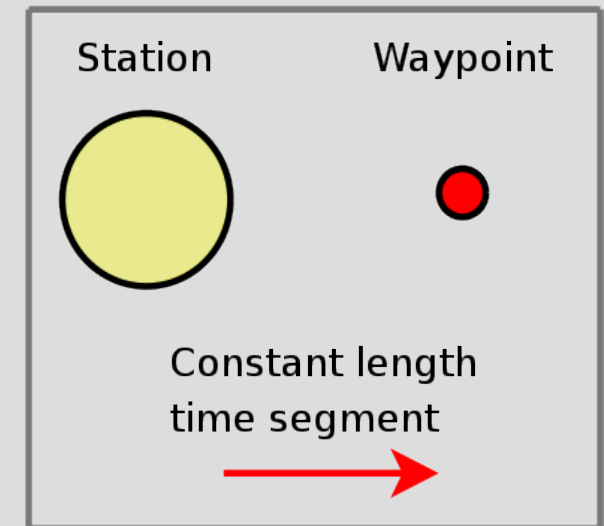
- Investigate optimal transfer strategies
 - Hub & spoke (e.g. bus feeders & light rail trunks)
 - Point-to-point (e.g. taxis, vanpools)
- Demand-responsive dynamic vehicle routing
 - Creates unique schedule based on demand inputs
 - Utilizes command, control, and monitoring networks
 - Emphasizes passenger service quality – high throughput, low latency, minimal vehicle movement
- Apply transit system constraints
 - Vehicle size (seating capacity)
 - Station size (berthing capacity)
 - Link connectivity (network topology)
- Multimodal layers of vehicles
 - various passenger capacities or network connectivity

Mass Transit Optimization Model Elements

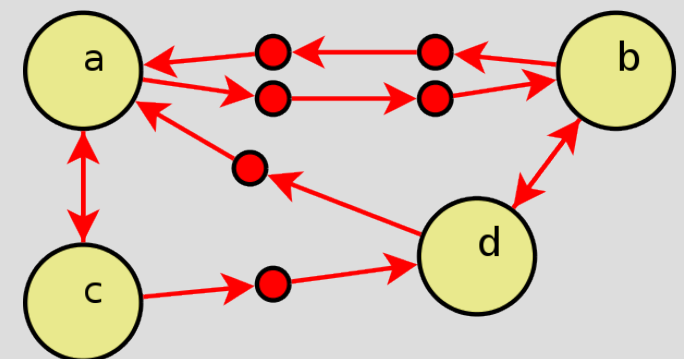
Modeled as an inventory problem

- **Station** nodes with quantities of passengers, vehicles
- **Links** between connected stations with quantities of passengers & vehicles in transit
- **Passengers:** grouped in bins by common current and final destinations
- **Vehicles:** multiple types with different capacities, station connectivity, and operating costs

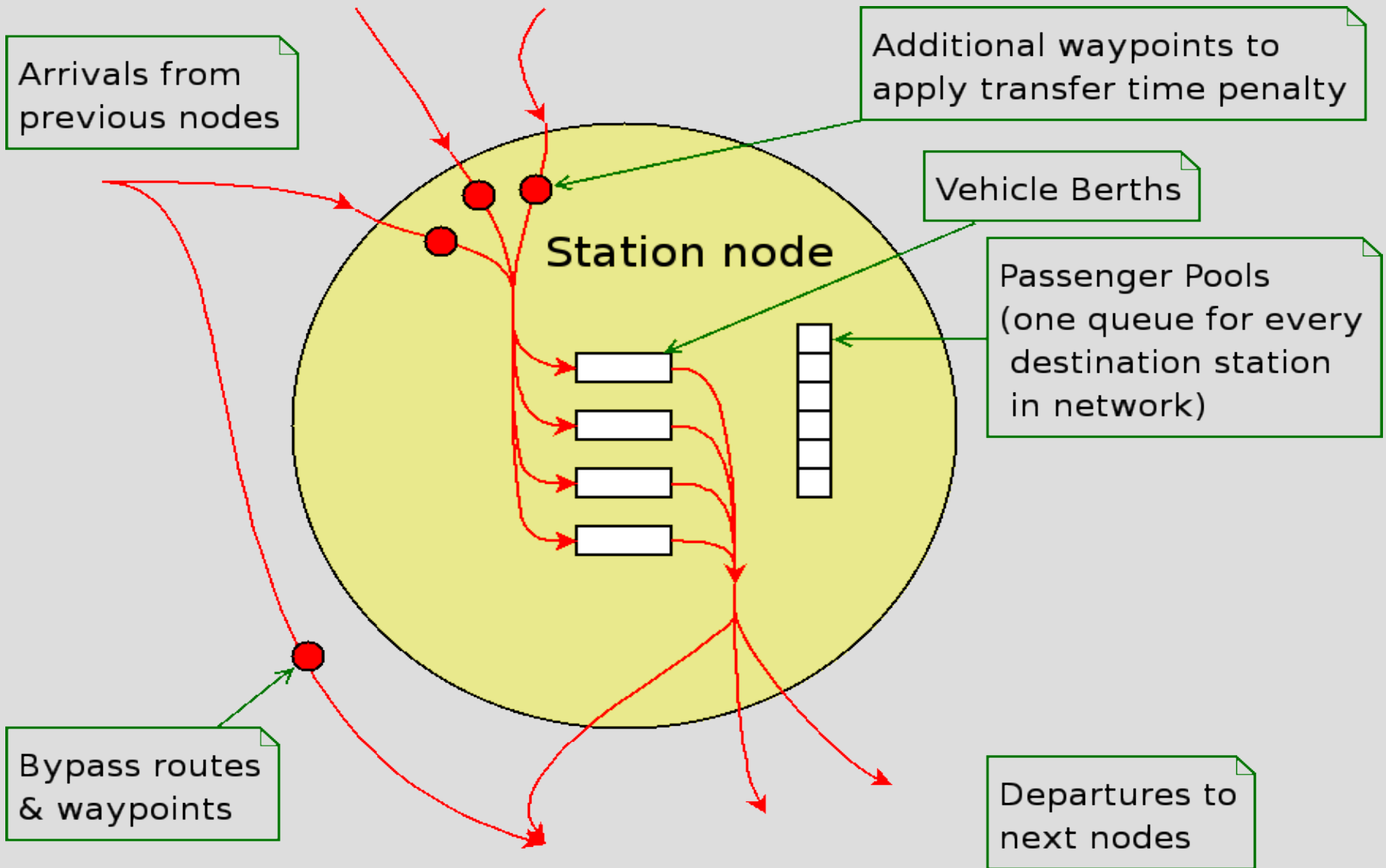
Model elements



Sample model of one transit layer

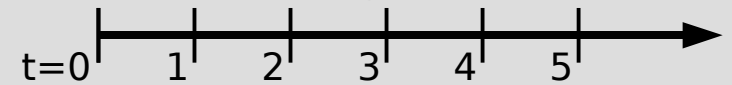


Conceptual Model of a Station



Transit Optimization Input / Output Variables

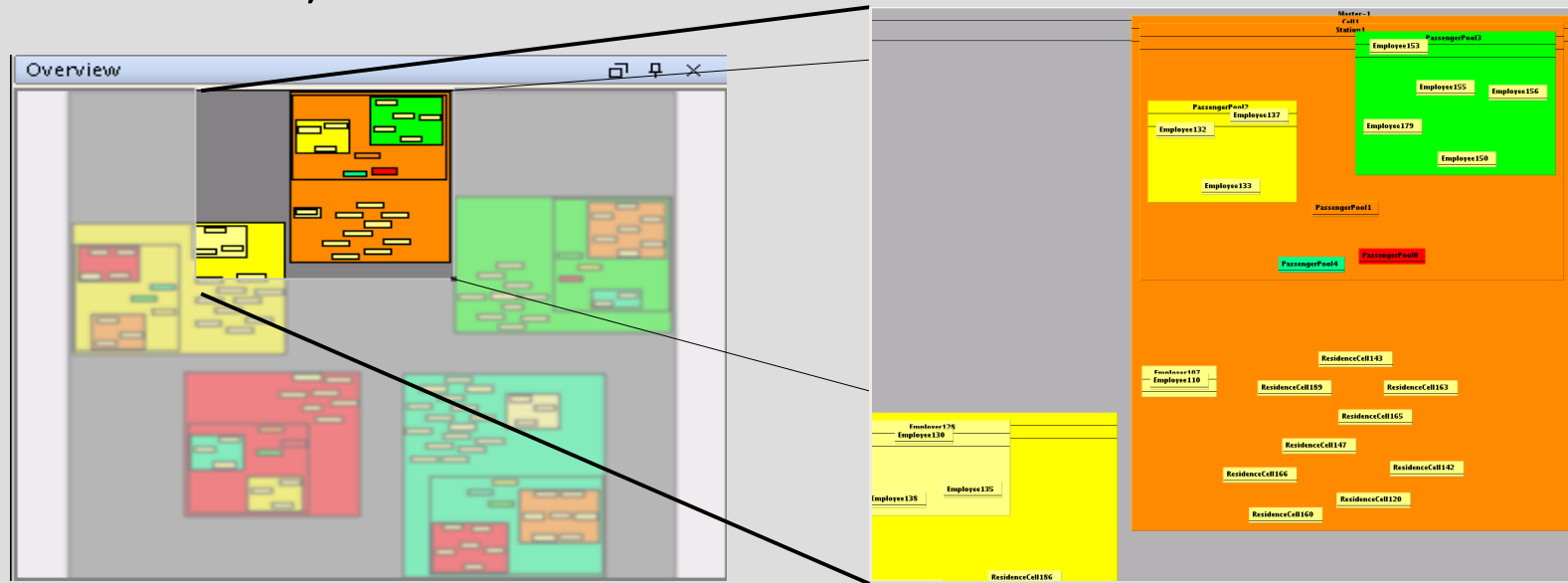
- Time represented by synchronous integer time steps



- Demand defined by initial passenger origins for each time step at each station

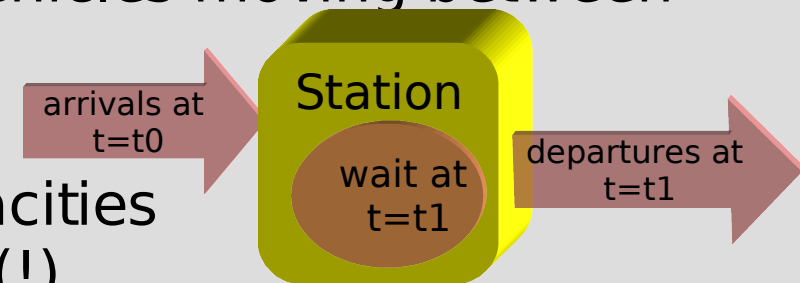
Output: schedule variables for each time step:

- Passenger locations, bulk movements
- Vehicle locations, bulk movements



Transit Optimization Constraints

- Inventory flow problem formulation:
 - Conservation of passengers & vehicles moving between nodes at each time step
- Passenger movement
 - constrained only by vehicle capacities
 - may transfer freely at any node (!)
- Vehicles constrained by:
 - connectivity matrix
 - station / waypoint node capacity
 - max fleet size limit



- Arbitrary constraints somewhat easy to add:
- e.g. “max vehicles on a link segment”
 - e.g. “max capacity on a group of waypoints”

Multiple Objectives

prioritized by weights:

Obj 1 >> *Obj 2* >> *Obj 3* >> *Obj 4*

1: Throughput

- Maximize passengers sent to final destination

2: Latency

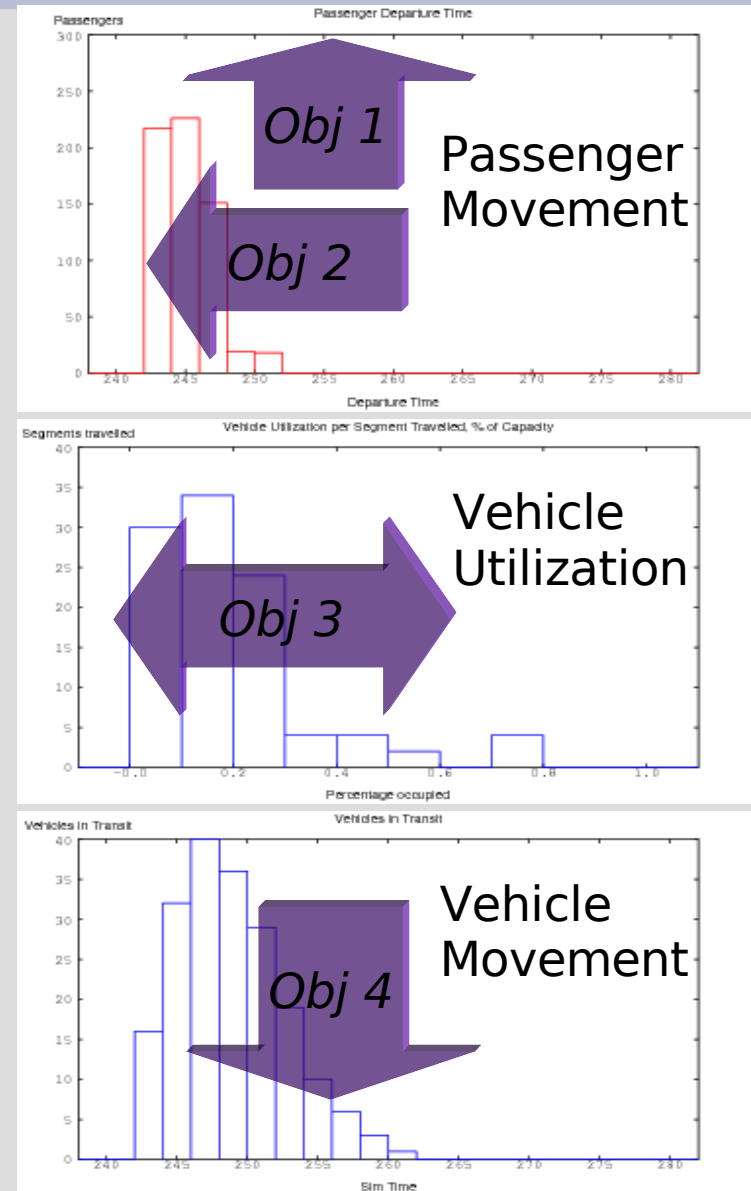
- Reward scheduler for delivering passengers earlier

3: Fleet Size *(Optional)*

- Minimize deviation from desired vehicle fleet size

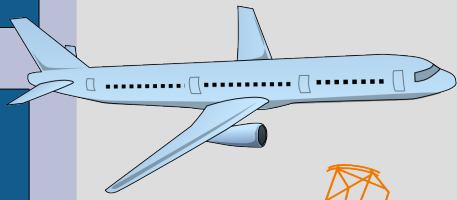
4: Operating Cost

- Minimize vehicle movements

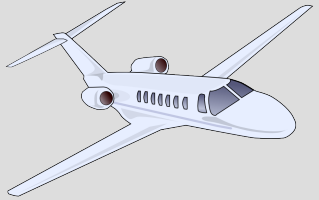


Transit Modes:

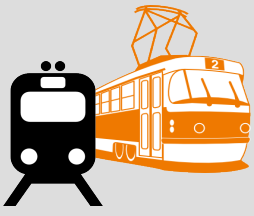
timing, capacity, and optimization parameters tuned to represent:



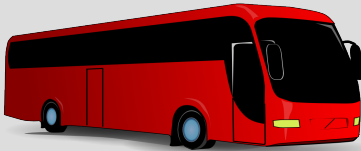
Aircraft



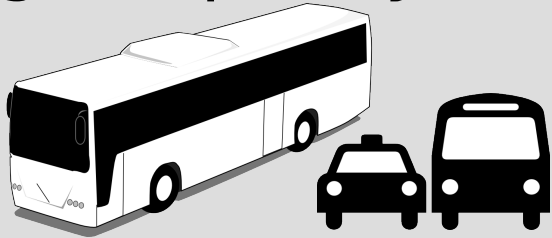
(original intent)



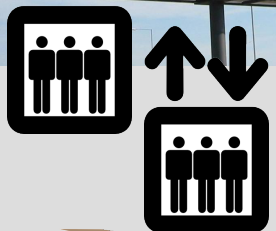
Subway / Rail (high capacity trunks)



Buses / Vanpools



Personal Rapid Transit networks



Elevators (!)



Automated Package Transport

Optimized Schedule Verified by Simulation

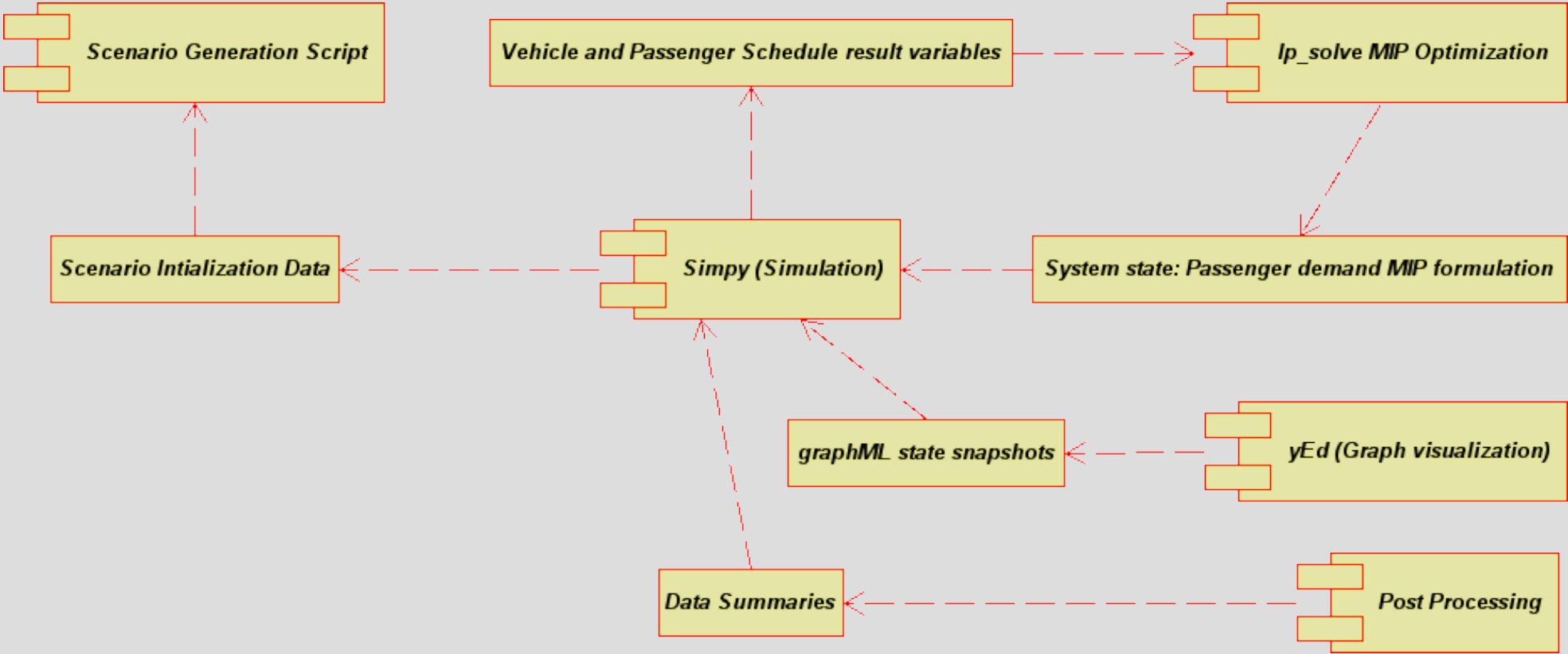
(the second half)

- Collects detailed performance metrics
 - Feasibility assurance
 - Continuous time execution of transit model based on integer time steps
 - Inspection & analysis of track logs from individual passengers and vehicles
- State persistence
 - Evolve system state with all known data
 - Reformulate and re-optimize schedule as scenario progresses and new input data is introduced
 - Eventually allow rolling horizon scheduling

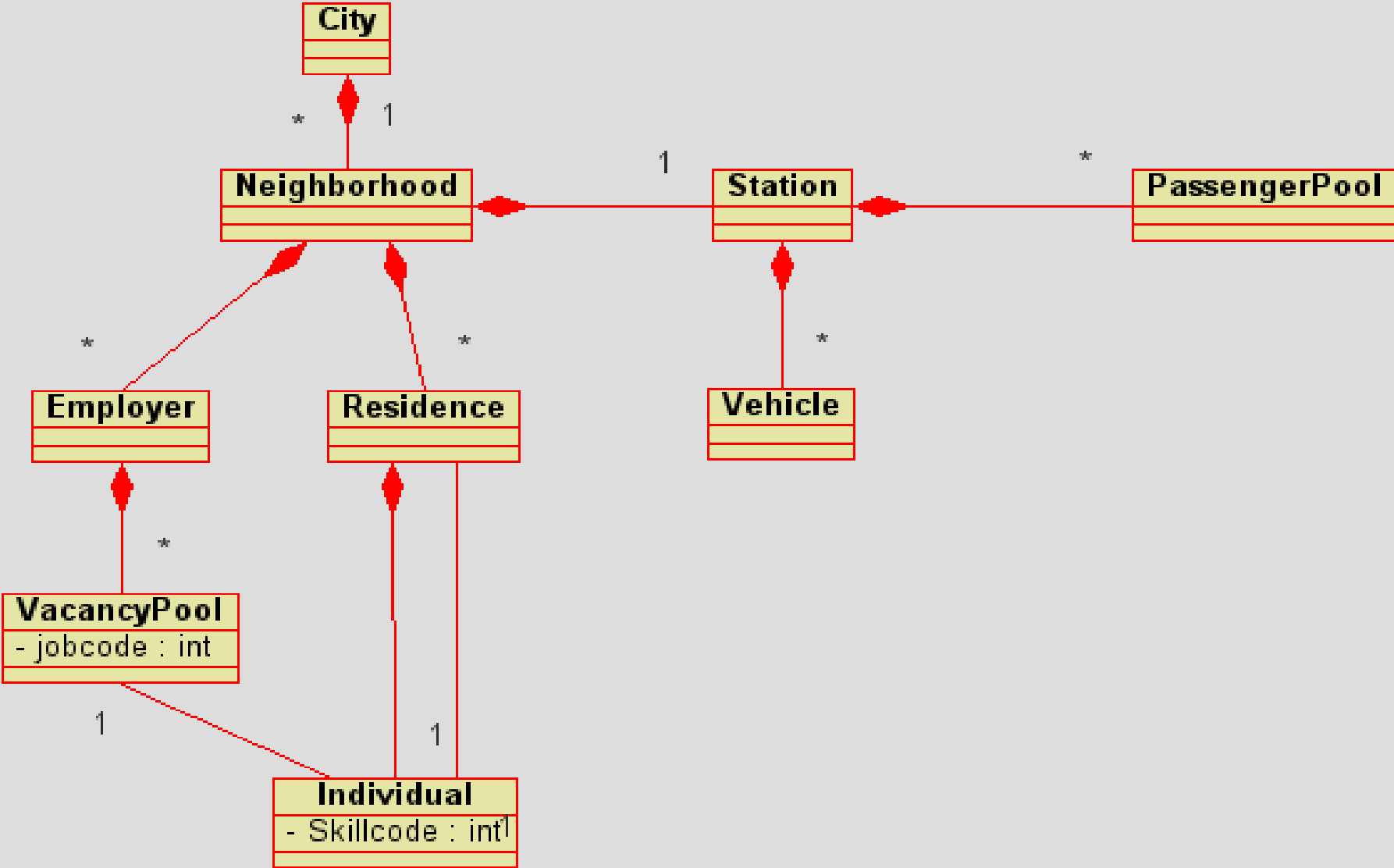
SimPy: discrete event simulation framework

LP_solve: MIP Optimization

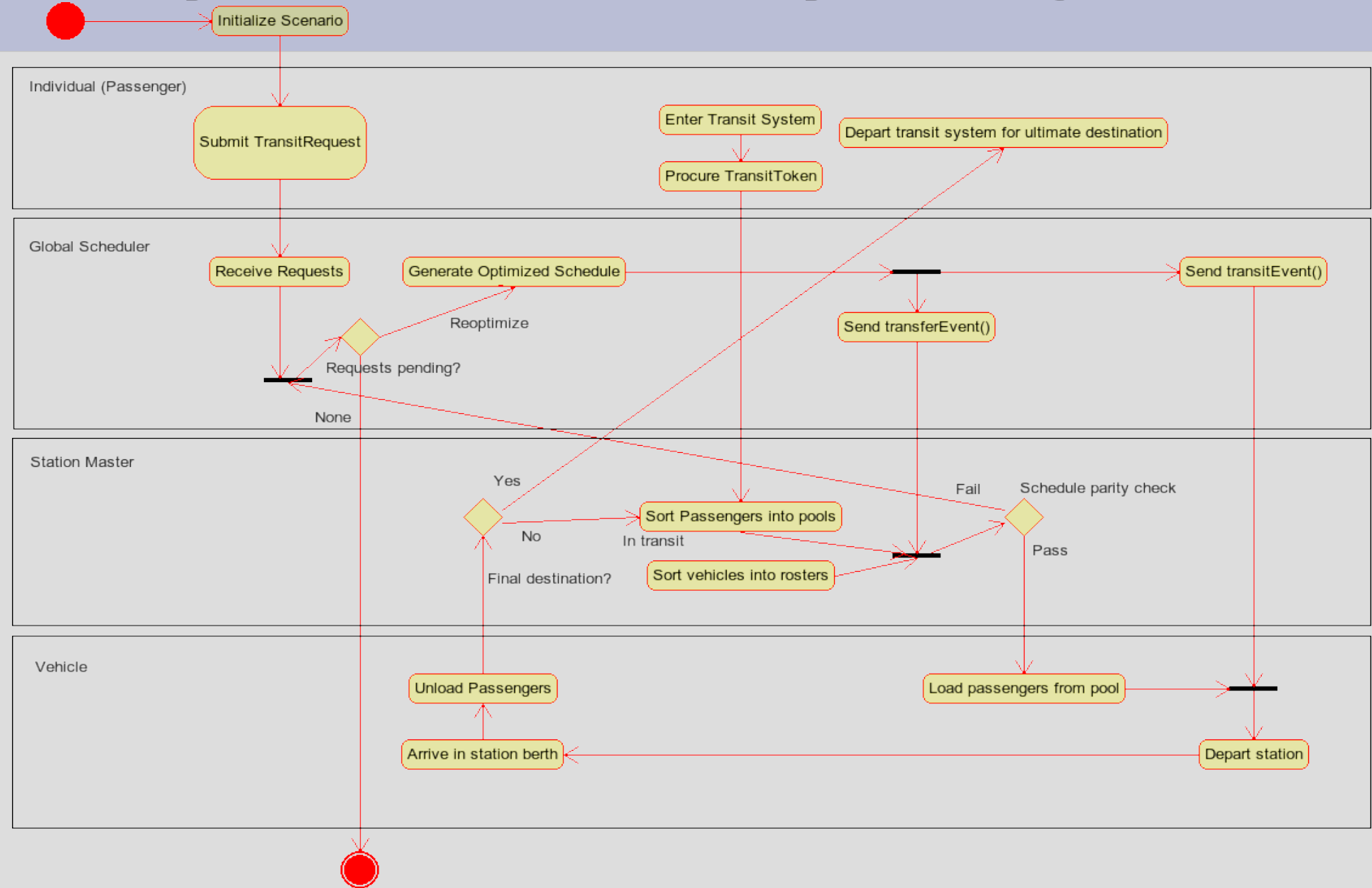
Simulation Component Diagram



Commuter Transit Model Class Structure



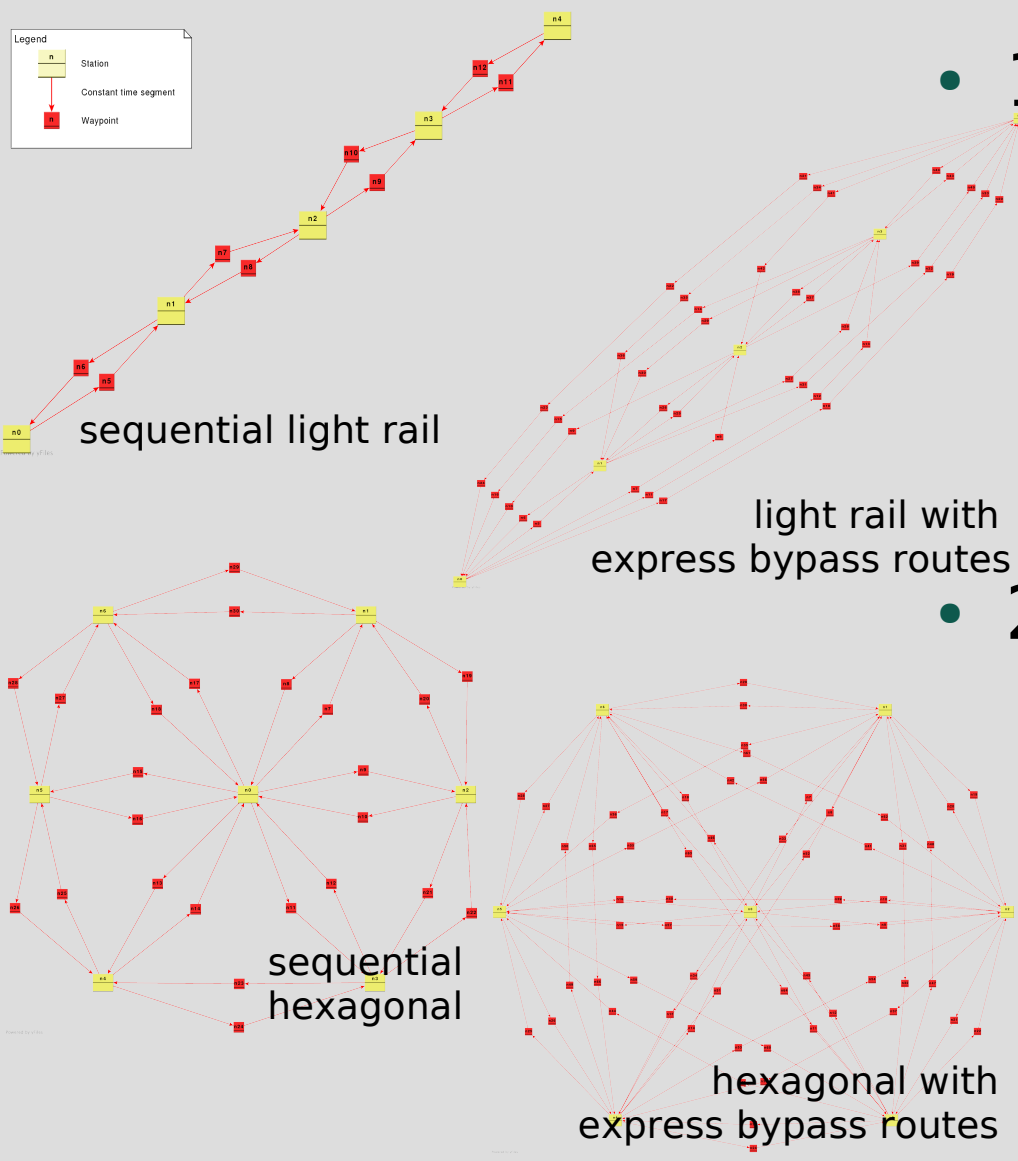
Commuter Transit Model System Activity Diagram



Verification and Validation

- Scenario Generation
 - Transit graph
- Demand Generation
 - Initial State
- Schedule Generation
 - MIP formulation: python code generates Ip model
- Schedule Results
 - Solution variables returned
 - Spreadsheet view
- Simulation of Results
 - Final state
 - Inspect individual passenger and vehicle histories

Parametric Analysis Scenarios



- **1D Light rail scenario**
 - extreme linear topology
 - with and without express routing (station bypass)
 - 7 station nodes
- **2D Hexagonal network**
 - extreme fully-connected star topology
 - with and without express routing (station bypass)
 - 7 station nodes

1D Rail Passenger Metrics

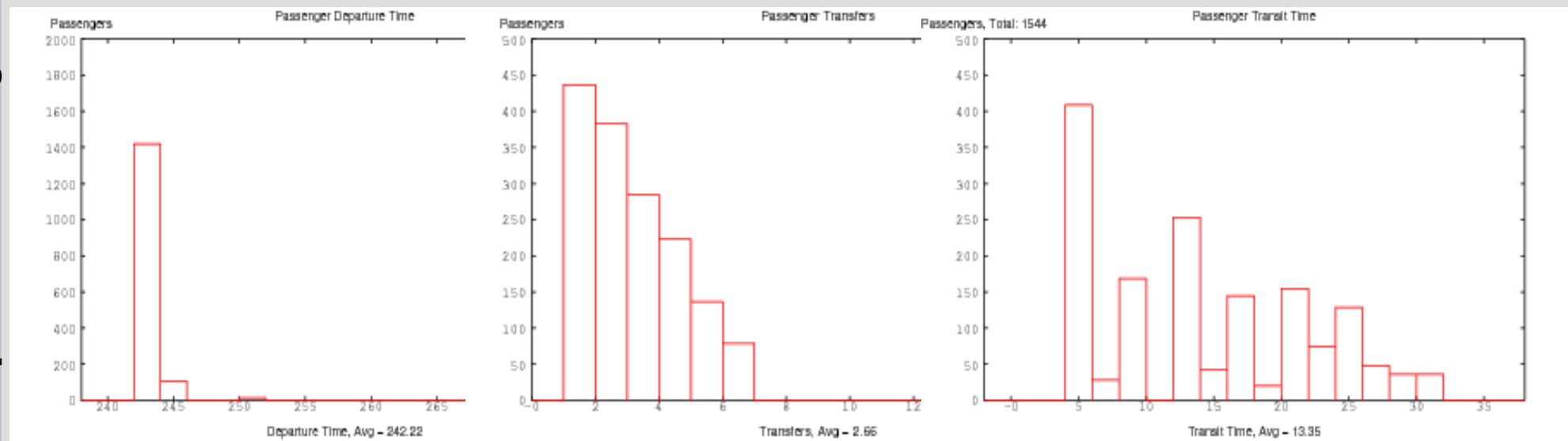
Response to uniform random demand pulse

waiting time (latency)

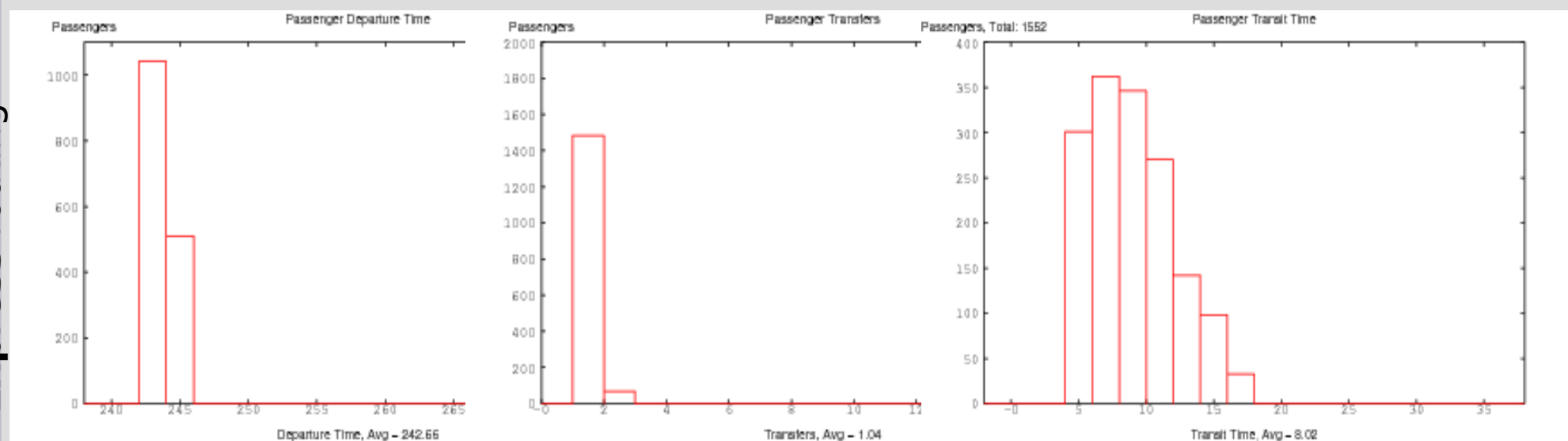
transfer stops (convenience)

travel time

Sequential routing



Express routing



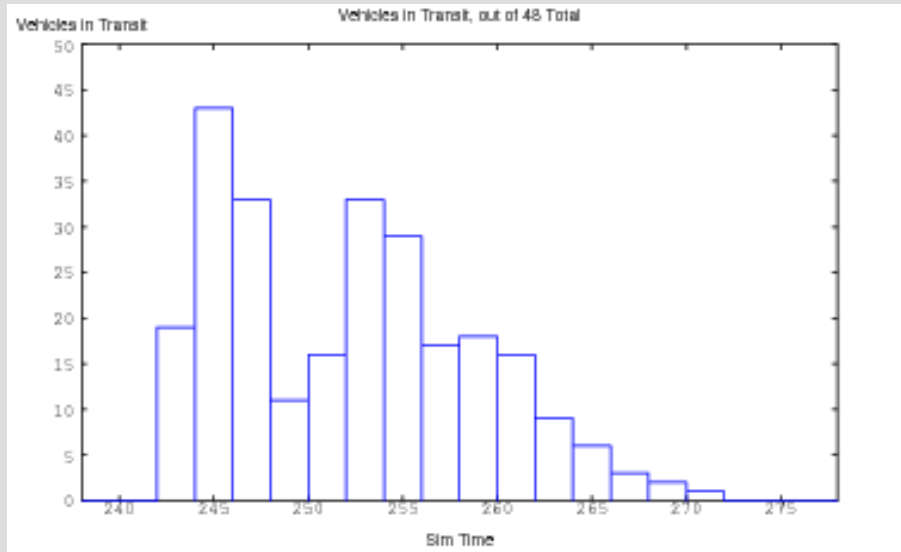
1D Rail Vehicle Metrics

Operating cost & efficiency

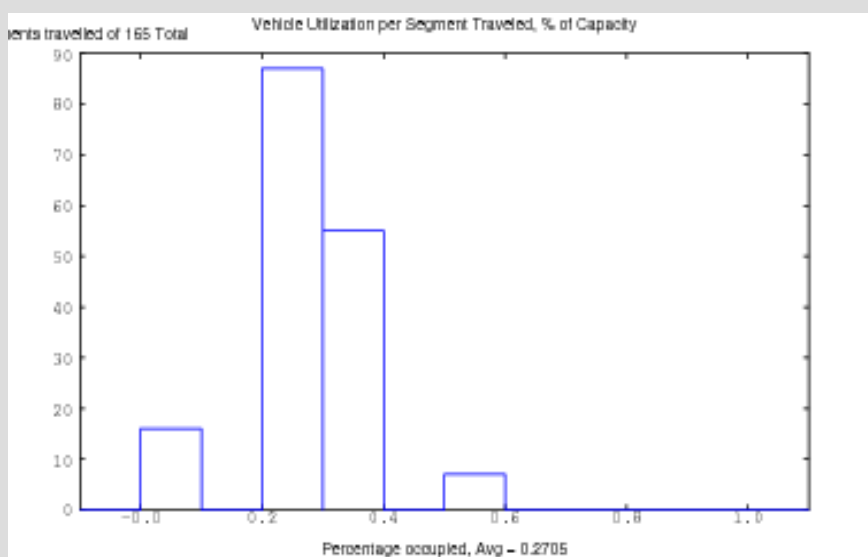
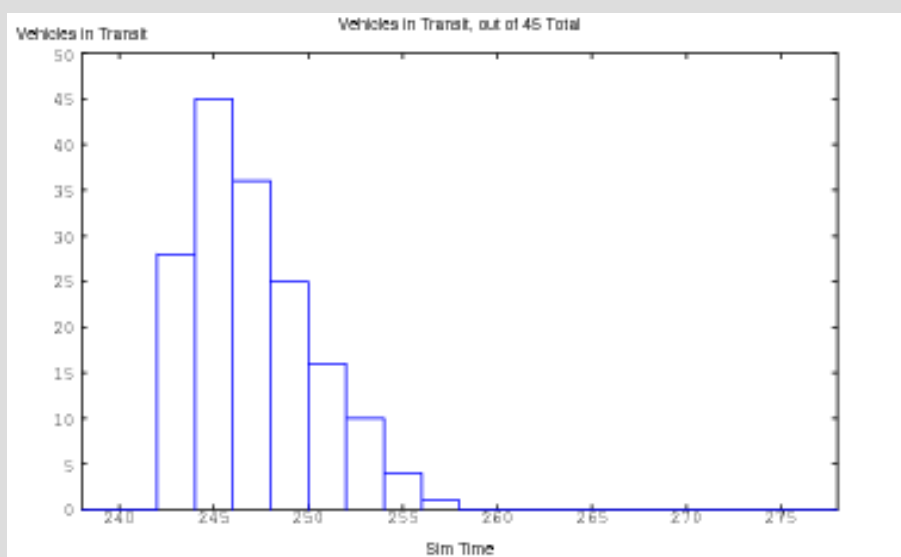
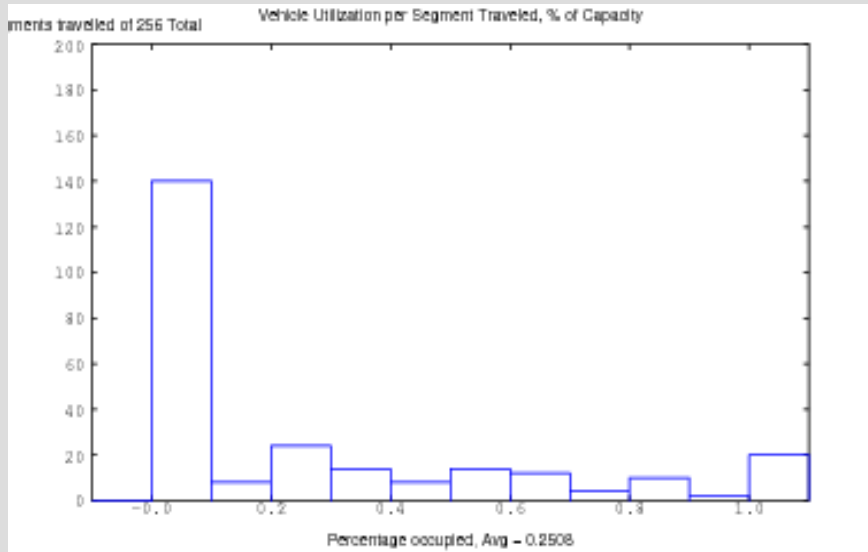
Sequential routing

Express routing

Vehicles in operation



Vehicle Utilization

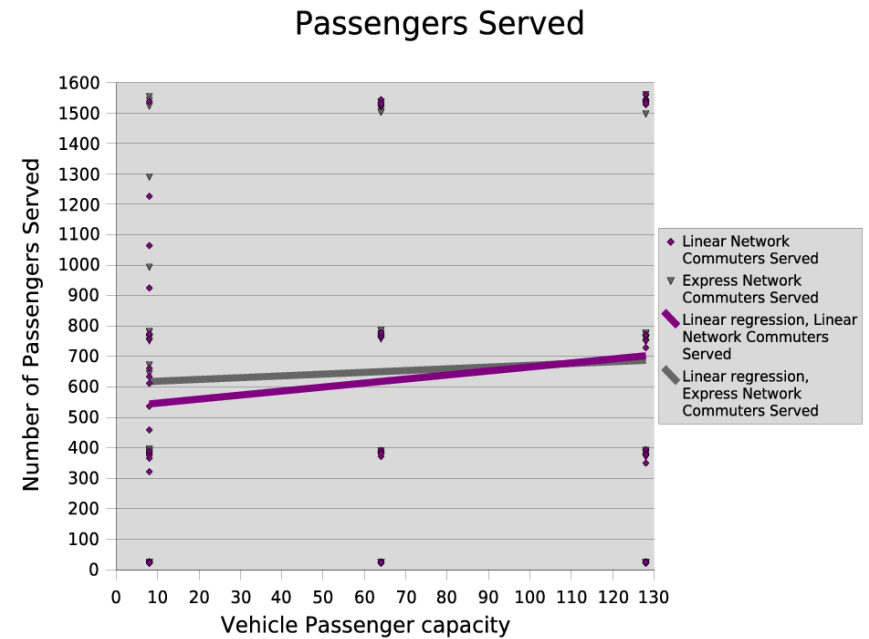
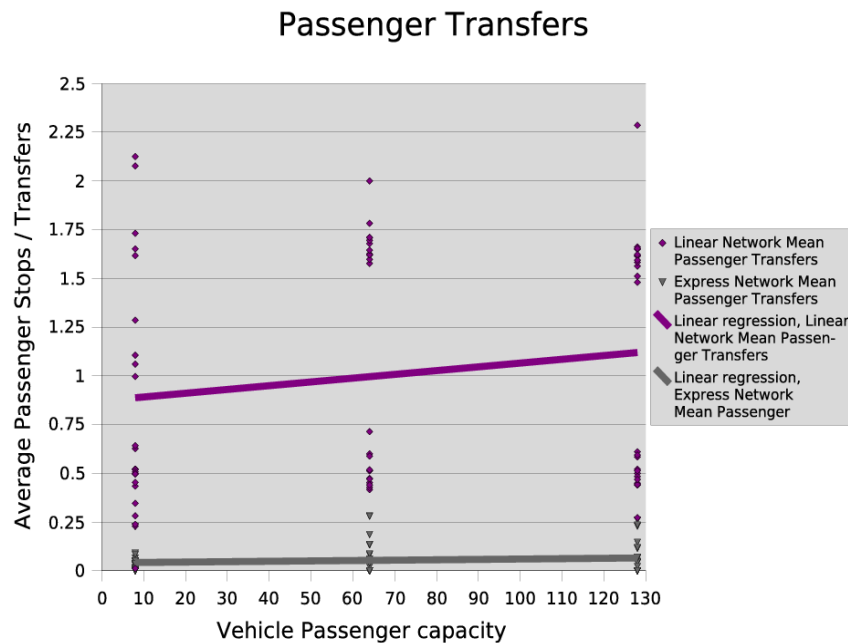
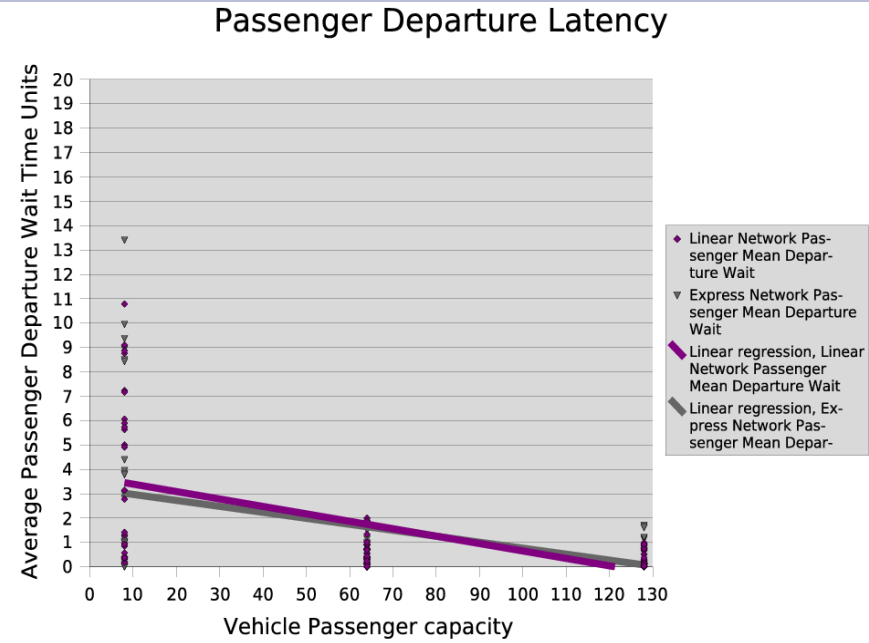
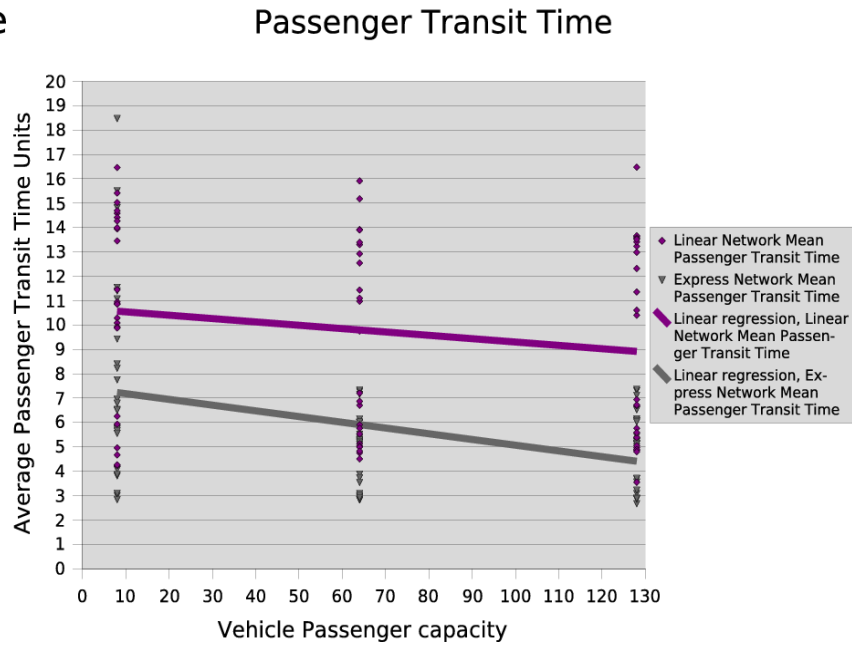


Factorial Experiments Design

- Design Parameters
 - Topology [linear **1D** Rail, **2D** hexagonal]
 - Offline stations [**sequential** routing, **express** routing]
 - Load per station [**4, 64, 128, 256**] commuters
 - uniform random distribution among origin stations
 - Vehicle size [**8,64,128**] passengers
 - Berths per station [**2,4,8**] vehicles
- Assumptions
 - Headways: 2 minute travel time across segments, 2 minute time to stop and transfer at a station
 - Impulse demand at $t = 240$ min
 - Vehicles must return to start configuration
 - Suboptimal & nondeterministic optimization timeout at 2 hours

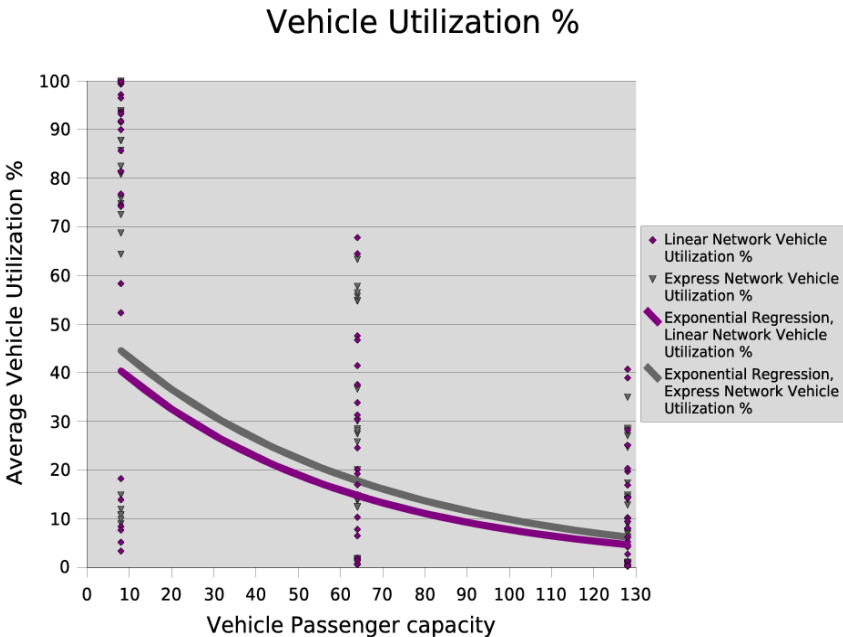
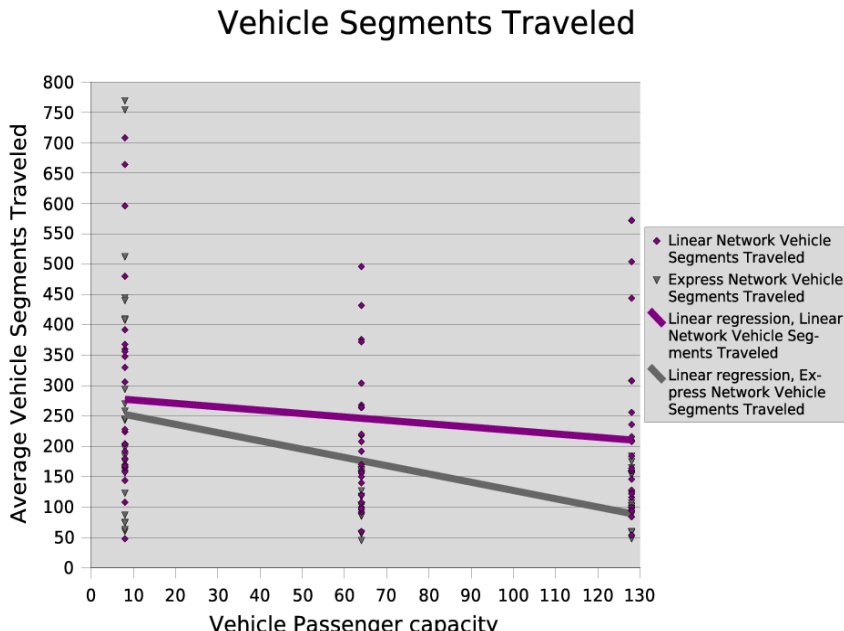
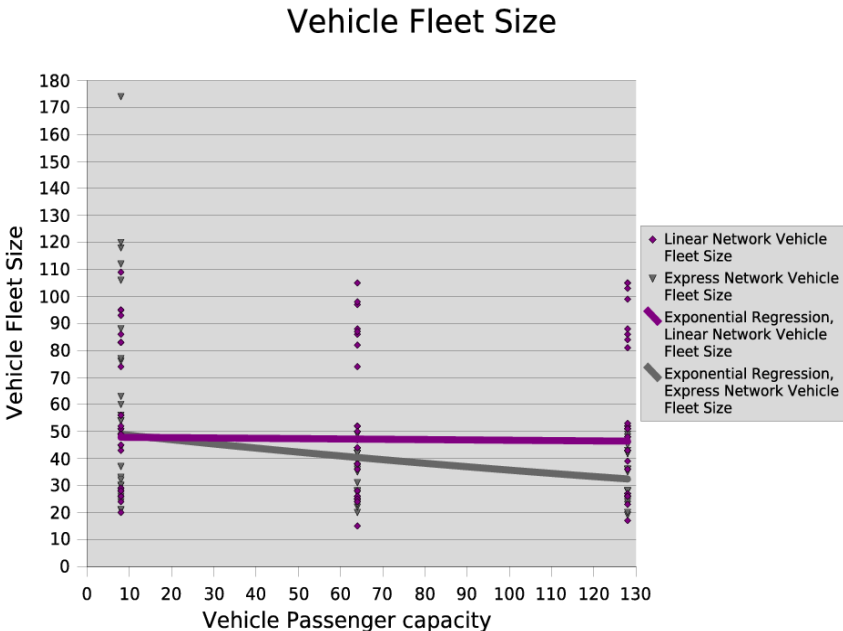
Passenger view of Sequential vs. Express routing with respect to Vehicle Capacity

Effect of Vehicle Capacity on Passenger Metrics



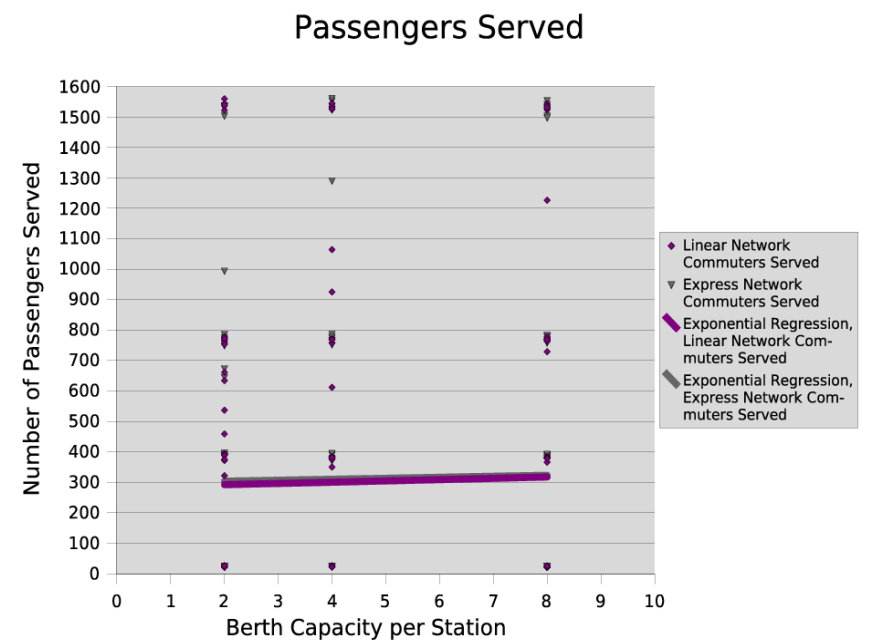
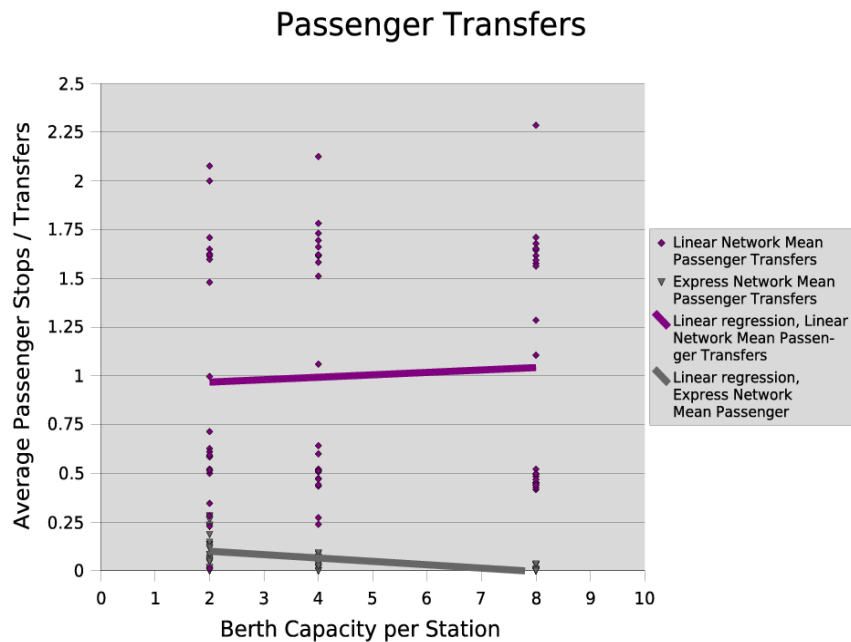
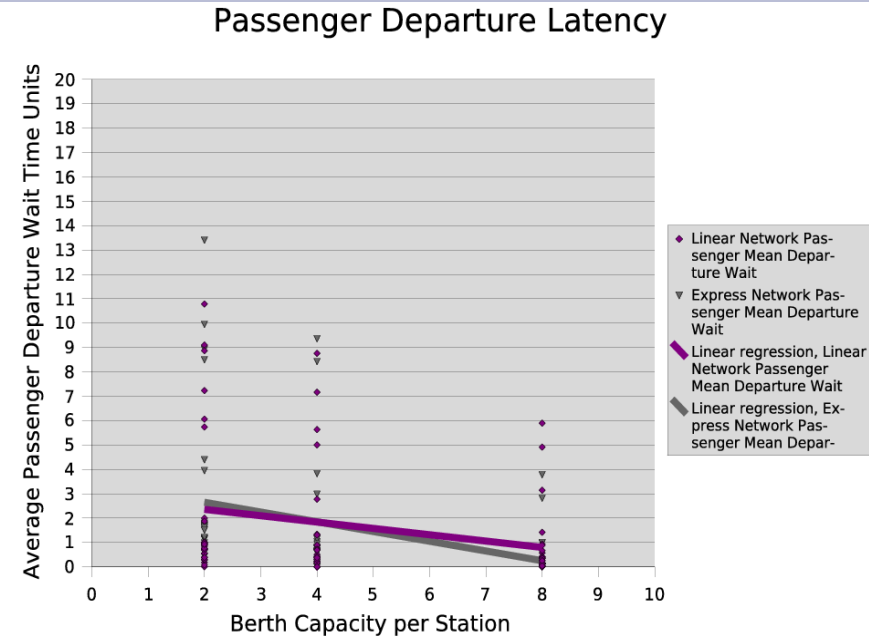
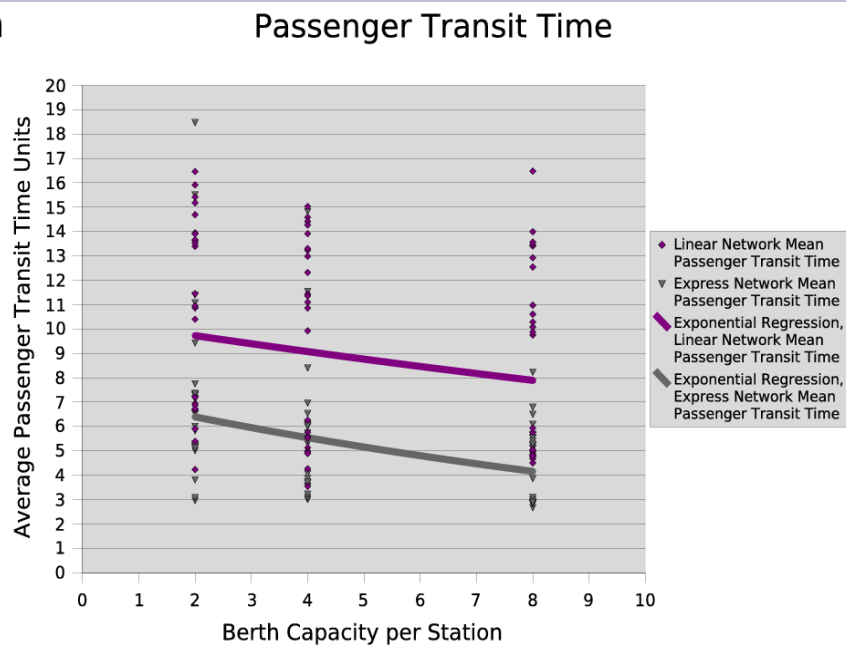
Fleet Operator view of Sequential vs. Express routing with respect to Vehicle Capacity

Effect of Vehicle Capacity on Vehicle Fleet Metrics



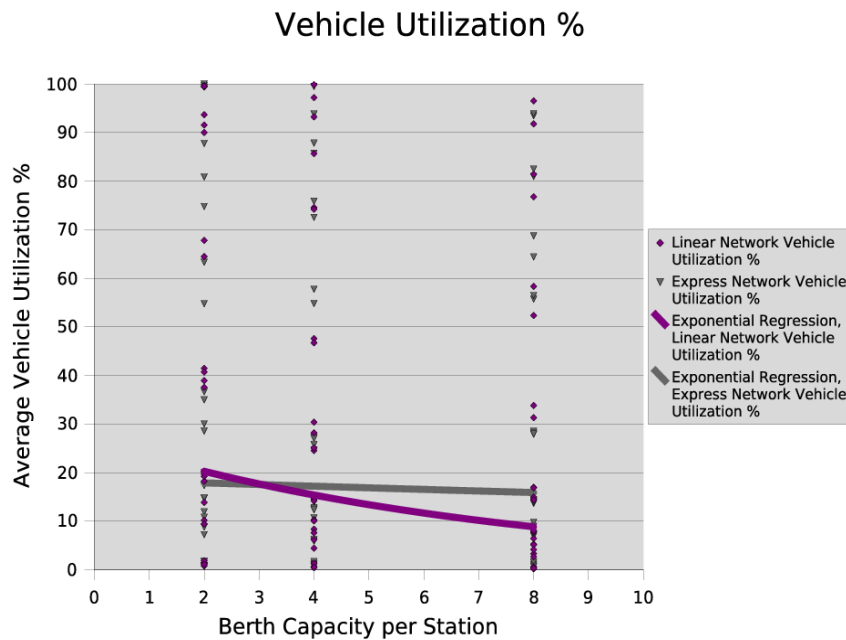
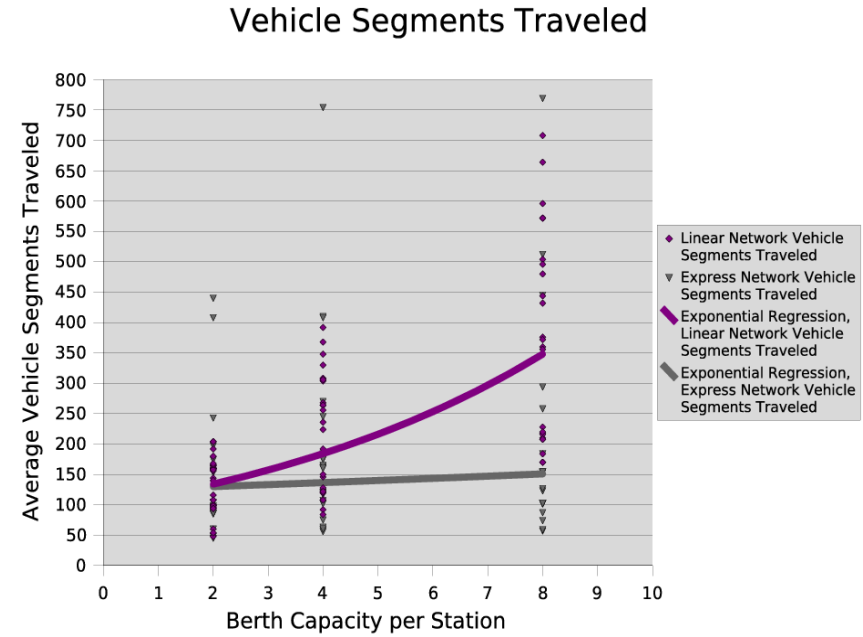
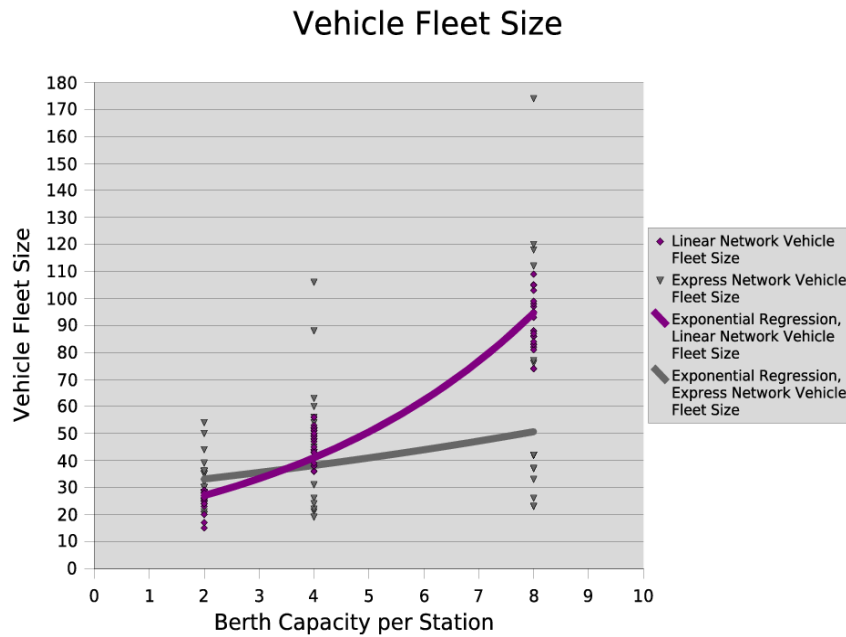
Passenger view of Sequential vs. Express routing with respect to Station Berth Capacity

Effect of Station Size on Passenger Metrics



Fleet Operator view of Sequential vs. Express routing with respect to Station Berth Capacity

Effect of Station Size on Vehicle Fleet Metrics



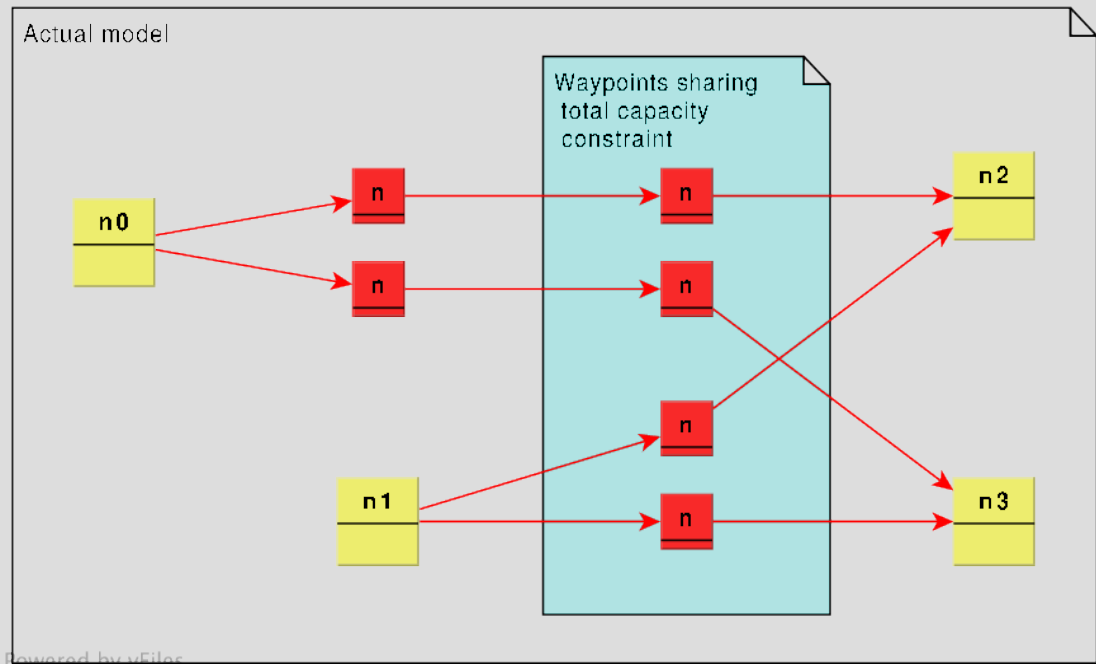
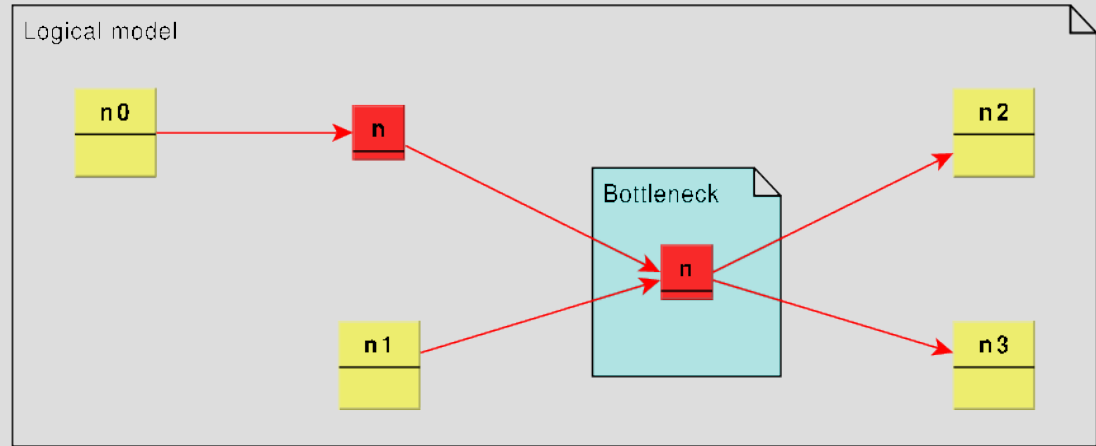
Conclusion:

This tool can do interesting things

- Dramatic improvement in mass transit performance possible by:
 - Using demand-responsive routing optimization
 - Constructing transfer stations off-line
- We can make mass transit perform as well as personally-owned vehicles
 - But this comes at a cost
 - Design transit-oriented development to keep network utilization at sustainable levels
- Analysts might use this tool to generate interesting data for trade studies

Future Work: Model feature completion

- State initialization to allow rolling time horizon
- Vehicle blocking on grouped constraints
- Priority passenger service via station queue manipulation

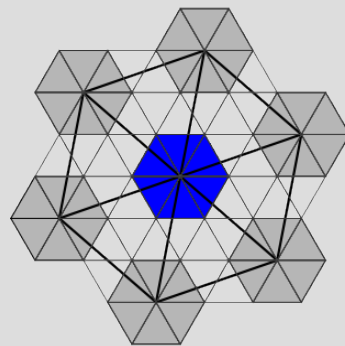


Future Work: Scalability

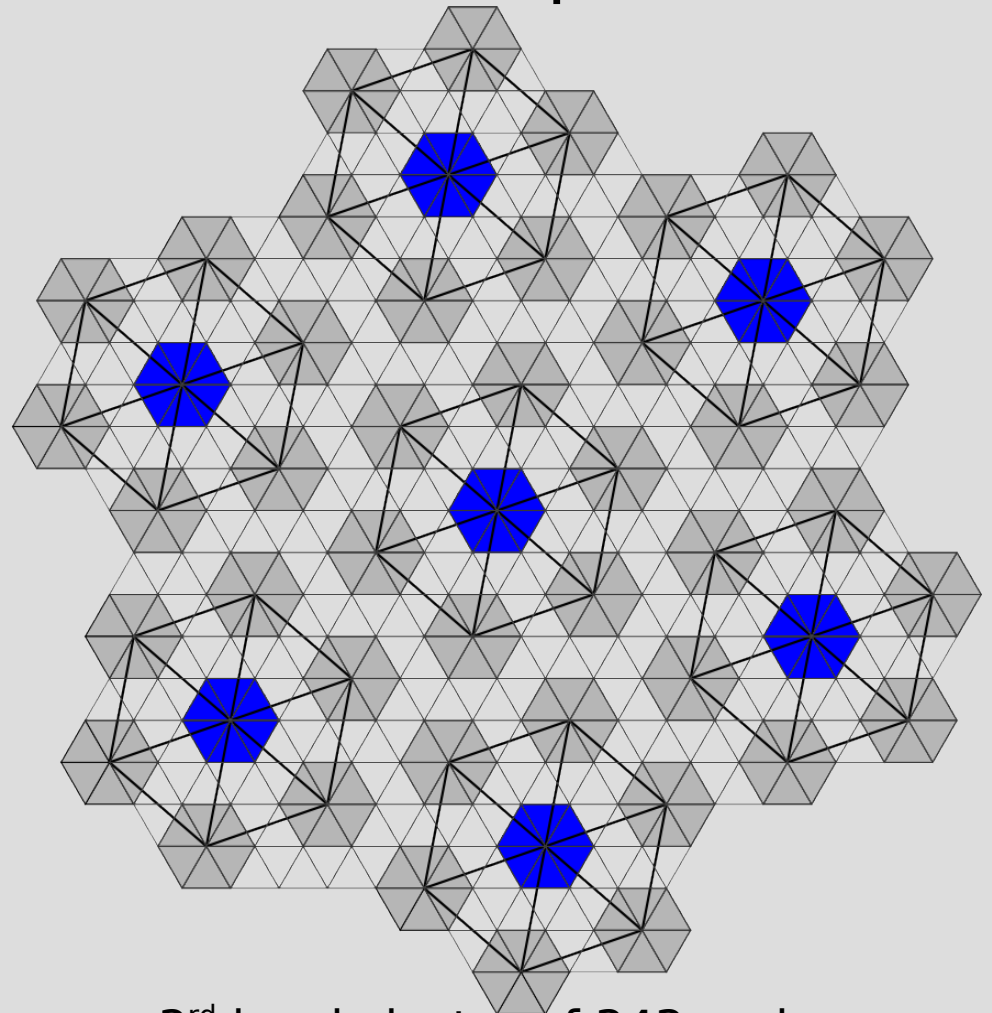
- Recursive Self-similar Hierarchical Space-Filling Structures



Basic 7-node unit



2nd level cluster of 49 nodes



3rd level cluster of 343 nodes

