

A real-time simulation-based decision support system for traffic management: the San Diego I-15 ICM model

IPAM Workshop IV: Decision Support for Traffic Matthew Juckes





Presentation Overview

- Program Background
- System Overview
- Network Prediction System (NPS)
- Response Plans & Corridor Scoring
- Maintenance & Real Time Targets
- Related Projects





FHWA ICM Program

7-Year Federally-sponsored program supporting development of improved mobility, safety, and other transportation objectives for people and goods through <u>integrated</u> <u>management</u> of transportation networks and cross-network connections in major transportation corridors in metropolitan areas.





FHWA ICM Program

- **Phase 1:** Conducted research into the current state of corridor management
- **Phase 2:** Develops analytic tools and methods that enable the implementation and evaluation of ICM strategies.
- Phase 3: Corridor Site Development, Analysis and Demonstration - San Diego and Dallas
- **Phase 4:** ICM Outreach and Knowledge and Technology Transfer



































Phase 3 – ICM Demonstration

- Five primary ICM goals:
 - 1. The corridor's multi-modal and smart-growth approach shall improve accessibility to travel options and attain an enhanced level of mobility for corridor travelers.
 - 2. The corridor's safety record shall be enhanced through an <u>integrated multimodal</u> <u>approach</u>.
 - 3. The corridor's travelers shall have the informational tools to make <u>smart travel</u> <u>choices within the corridor</u>.
 - 4. The corridor's institutional partners shall employ <u>an integrated approach</u> through a corridor-wide perspective to resolve problems.
 - 5. The corridor's <u>networks shall be managed holistically</u> under both normal operating and incident/event conditions in a collaborative and coordinated way.





ICMS Innovation

Paradigm shift to more pro-active traffic management method, making use of prediction tools, on-line micro simulation and improved decision support







Location Description

- 21 mile North-South Stretch of I-15 between SR-78 and SR-163
- I-15 is a 6-10 lane freeway with a 4 lane managed HOV/SOV toll road in the center
- Multiple parallel arterials including: Black Mountain Road, Pomerado Rd and Center City Parkway
- Encompasses three cities: Escondido, Poway, San Diego
- A microsimulation network coded in Aimsun software
- 60 minute prediction engine coded in Aimsun Online







ICM System





Model Description

I-15 ICM Model generated from the SANDAG Regional Macro Model





Data Inputs and Outputs





Data Inputs and Outputs





System Evolution

Response Plan Evaluation





Multi



Aimsun Online Architecture





DSS Model Flow











Network Prediction System

• The Network Prediction System (NPS), summarizing, consists in an analytical model for each detector such that:

$$Y_D^{k,t} = f(X^t),$$

where

- $Y_D^{k,t}$, represents the k predicted variable for detector D at time t.
- *f* , represents the analytical model.
- X^t , represents the real detector measures available at time t.





NPS – Problem Description

• "Given a time point t and a detector D, a prediction of the flow (or other provided measure) of that detector at time t+h is wanted, using all the information available at the moment t". Mathematically expressed:

$$Y = f(X)$$

where

- Y is the objective variable (flow, occupancy or speed at time t + h),
- X is the matrix containing all the necessary input information at time t,
- *f* is the model representing the relationship between *X* and *Y*.
- The matrix X is built with the explicative variables of the model, that is those variables considered as necessary to explain the behavior of Y, including:
 - Previous values of the selected detector *D* at time points *t*, *t*-1, *t*-2 ...
 - Previous values of the upstream and downstream of D at time points t, t-1, t-2 ...
 - Or even other quantities such as occupancies or speeds. The rows of *X* determine the observed day.
- The model *f* has to be trained with historical data and has to be estimated by means of statistical techniques. The set of upstream and downstream have also to be specified.





NPS – Problem Description







NPS – Technical Issues

- Linearity
- Multicollinearity
- Variable selection/Dimensionality Reduction
- Missing Data





NPS - Methodology

- LASSO (least absolute shrinkage and selection operator) is an alternative regularized version of least squares which uses the constraint that ||·||₁, the L₁-norm of the parameter vector, is no greater than a given value.
- Regularization methods are used for model selection, in particular to prevent overfitting by penalizing models with extreme parameter values. One of the most common variants in machine learning is L_1 -norm (being the other the squared L_2 -norm) regularization, which can be added to learning algorithms that minimize a loss function E(X, Y) by instead minimizing

$$E(X,Y) + \alpha ||w||,$$

where

- w is the model's parameter vector,
- $\|\cdot\|$ is the L_1 -norm,
- α is a free hyperparameter that needs to be tuned empirically (in ours case using the described k-fold cross-validation).





NPS - Methodology

- L₁-norm regularization is often preferred because it produces sparse models and thus performs feature selection within the learning algorithm (increasing the penalty cause more and more of the parameters to be driven to zero and this deselects the features from the regression, thus Lasso automatically selects more relevant features and discards the others).
- But since the L_1 -norm is not differentiable, it may require changes to learning algorithms, in particular gradient-based learners. For that reason the optimization problem may be solved using quadratic programming or more general convex optimization methods, as well as by specific algorithms such as the least angle regression algorithm.



NPS – Accuracy and Validation

- When training a model, the deviations of the predictions from the data set used for training (such as the R² coefficient in linear regression) can be calculated. However, these errors are not representative of the error within the model when another set of new data is used. Thus, another way to measure the error is needed.
- In order to select the model with higher accuracy, a *k*-fold cross-validation scheme is used. This is the most standard way to ensure that the errors estimated in the training stage will be consistent with future predictions. In the k-fold cross-validation, the set of historical days is split into k groups.
 For each group the procedure is:
 - 1. Remove the group of days (called the test sample) from the historical data. The remaining set is called the training sample.
 - 2. Train the model with the training sample.
 - 3. Evaluate predictions with the test sample
 - 4. Then the mean of the errors achieved by these k test samples is calculated.
- In our problem, k = 10.



NPS – Accuracy and Validation

- Two performance evaluation measurements have been used.
 - Mean absolute percentage error (MAPE) is defined:

$$MAPE = \frac{1}{n} \sum_{t=1}^{n} \left| \frac{y_t - \hat{y}_t}{y_t} \right|$$

where

- y_t is the actual value,
- \hat{y}_t is the forecast value.
- Quality measurement criteria:
 - If flow > 750 veh/h: check if relative error (MAPE) is below 15% to satisfy the quality criteria
 - Else: check if absolute difference between actual and forecast is below 150 veh/h.





(aggregated results)

Forecast horizon [min]	Minimum error [%]	Maximum error [%]	Mean error [%]	Standard deviation [%]
15	2.43	23.75	7.87	4.49
30	2.8	24.7	8.43	4.49
45	3.06	24.67	8.72	4.51
60	3.23	25.78	8.99	4.56
75	3.37	26.47	9.24	4.61

Forecast horizon [min]		Maximum Quality ratio [0, 1]	Mean Quality ratio [0, 1]	Standard deviation [0, 1]
15	0.93	1	1	0.01
30	0.89	1	1	0.01
45	0.85	1	0.99	0.01
60	0.83	1	0.99	0.01
75	0.81	1	0.99	0.02



NPS – Tests with Real Data





NPS – Tests with Real Data







Pattern Calibration





- 11 Day Patterns
 - Weekdays
 - Holidays
 - Rainy Days
 - Special events

Aimsun (Tasks Logs Executors Dashboard Logged in as anonymous Controller No running tasks USER ACTION Uptime: 42:00:05 No queuing tasks GL icenses IF Help KII → Quit Sensor Server Patterns: 11 Untime: 27:05:03 Real Time: Mon Jun 02 2014 06:59:00 GMT-0700 (Pacific Davlight Time) Forecasted: Mon Jun 02 2014 08:10:00 GMT-0700 (Pacific Daylight Time) Name: ICMSAIM07 Kill More Info Address: 10.75.1.71 OS: Windows Version: > 7 Map Server Model UID: {0ddf7f6e-e3e1-42dd-81b8-a4547c3079583 Rite: 64 Uptime: 41:59:51 Model Name: Project Cores: 32 Model Location: //10.75.1.58/OnlineData/AOData/SIM/TSS_SANDAG_NETWORK_V8.ang RAM: 128GB Kill Model DB: QMYSQL://root@10.75.1.58:3306/sandag View Man EXECUTORS 10.75.1.59 Pattern Matcher Current Pattern: 1 monday Uptime: 42:00:05 10.75.1.60 10.75.1.61 Kill 10 75 1 62 10.75.1.70 APM Server Running an Analytical + Simulation process for Mon Jun 02 2014 06:55:00 GMT-0700 (Pacific Dayligh Uptime: 42:00:05 Time) More Info Kill Prediction vs Real: Mon Jun 02 2014 06:50:00 GMT-0700 (Pacific Daylight Time) QM: 79.014991% Quality Server

Aimsun Online Dashboard

- 15 Minute data sets
- Runs every 5 minutes



Corridor Performance Needs



Measures

- Intersection
- Ramp Meter
- Express Lanes
- Sections
- Transit
- Routes
- Targets (within 15%)
 - 0-15 minutes 92%
 - 15-30 minutes 80%
 - 30-60 minutes 40%





SIMULATION SYSTEMS

Response Plan Scoring





ATMS - Internet Explorer



- - X

ATMS - Internet Explorer

Attps://www.icmsd.net/atms/atmslite.jsp



ATMS Login - Intern...

Desktop 🐣 🔺 📑 🛱 👘 8:21 AM

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San Diego 511 - APP





●●ooo AT&T LTE Person	10:54 AM al Hotspot: 1 Connec	- ≁ ∦ 76% □= etion
I-15 NB (ROUTE 16	63 TO SR-78)	
Now		18m
At 11:00 AM		18m
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Last Updated: 10:5	50 AM	
I-15 SB (SR-78 TO	ROUTE 163)	
Now		21m
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At 11:15 AM		26m
At 11:30 AM		24m
At 11:45 AM		21m
Last Updated: 10:5	50 AM	

Transit Travel Times





System Configuration

Save Cancel

Corridor Global Asset Availability	Strated	v Matrix Sim C	Corridor Sco	ore Transit		
Response Plan Improvement		<mark>8.0</mark> %				
Evaluation Area		Conservative	Moderate	Aggressive		
"Upstream 3x/Downstream 1x Length Bound"		0.5 mi	1.0 mi	<mark>5.0</mark> mi		
"Lateral Width Bound"		1.0 mi	2.0 mi	3.0 mi		
Vehicle Occupancy Factors		sov	ΗΟΥ	Truck	Bus	
		1.0	2.3	1.5	25.0	
Evaluation Interval Factors		15 Min	30 Min	45 Min	60 Min	
		0.20	0.40	0.20	0.10	
Include LOS Processing?		⊻				
LOS Comparison		30 Min				
LOS Exclusions		A	🗹 в	C	D	E

Corridor 💽 🛛 RM Sim Hierarchy 🛛 Express Lanes Response 🖉 Congestion Event Finder 🖉 Congestion Thresholds 💽 👁 Arterial Freeway Congested Congestion Congestion Events Events V/C 4 Speed Minimum Speed Differential vs Free Flow 15.0 mph 1.0 mph 1.0 % Minimum % Speed Change vs Free Flow 25.0 % 30.0 sec/veh 30.0 sec/veh Minimum Delay Differential vs Free Flow Minimum V/C Differential vs Free Flow 0.70 0.80 Freeway Arterial **Over Congested** Congestion Congestion Events Events Speed Delay \bigtriangledown Minimum % Change in Speed vs Historical 20.0 % 1.0 % Minimum Speed Differential vs Historical (-) 10.0 1.0 Minimum % Change in Delay vs Historical 1.0 % 20.0 % Minimum Delay Differential vs Historical (-) 20.0 60.0

Save Cancel




 Bounding Box – size defined by response posture to identify minimum selection set

2. LOS Comparison – to identify additional links where response plan implementation has changed operations within the network





LOS Comparison

Link	Do Nothing	Response Plan A	Response Plan B	Response Plan C	Included in Evaluation?
1	С	D	E	С	Yes
2	В	А	В	В	No*
3	С	С	С	D	Yes
4	С	С	С	С	No
5	F	F	F	F	No
6	D	E	D	С	Yes
7	С	С	С	В	Yes

* This link is not included with LOS A and LOS B exclusion options turned on.

Determined by comparing each Response Plan against 'Do Nothing'

Corridor Score =
$$\frac{D_0 - D_z}{D_0} * 100$$

- D_0 = Person-delay under Do Nothing Case
- z corresponds to the response plan evaluation
 - D₁ = Person-delay under Response Plan A
 - D₂ = Person-delay under Response Plan B
 - D₃ = Person-delay under Response Plan C



Person Delay per link = $(SV_{SOV} + HV_{HOV} + TV_{Truck} + BV_{Bus}) * D_{Avg}$

- D_{Avg} = Average Link Delay
- Configurable Static Occupancy Factors:
 - » S Single Occupancy Vehicle
 - » H High Occupancy Vehicle
 - » T Truck
 - » B Bus
- V_{SOV} = Link Volume of SOVs
- V_{HOV} = Link Volume of HOVs
- V_{Truck} = Link Volume of Trucks
- V_{Bus} = Link Volume of Buses





- Person delay is totaled for all of the Evaluation Area links, for each Response Plan and the Do Nothing case. Resulting in four corridor delay values for each 15 minute time period.
 - D₁₅ = 0-15 minute
 - D₃₀ = 15-30 minute
 - D₄₅ = 30-45 minute
 - D₆₀ = 45-60 minute



The Corridor Person Delay value utilized for comparison of response plans is determined by factoring each of the 15 minute evaluation period by configurable weighting factors: $D = (W_{ex} * D_{ex}) + (W_{$

 $D_{y} = (W_{15} * D_{15}) + (W_{30} * D_{30}) + (W_{45} * D_{45}) + (W_{60} * D_{60})$

- $W_{15} = 0.15$ minute weighting factor
- $W_{30} = 15-30$ minute weighting factor
- $W_{45} = 30-45$ minute weighting factor
- $W_{60} = 45-60$ minute weighting factor
- $W_{15} + W_{30} + W_{45} + W_{60} = 1$

- y corresponds to the Response Plan Evaluation
 - \circ 0 = Do Nothing
 - 1 = Response Plan A
 - 2 = Response Plan B
 - ∘ Etc.



\bigcirc

Corridor Score

New Methodology Being Explored

- Travel Time Index & Travel Time Reliability
 - Percentage of trips below Travel Time threshold
 - Freeway Threshold example 1.33 * Free Flow Travel Time;
 - Arterial Threshold example 2.50 * Free Flow Travel Time;













Maintaining the Model

370000r

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Reasons for Model Updates

- Change in travel patterns and demands;
- New Infrastructure;
- New ITS systems;
- New Public Transit;
- New Developments;



Real Time Validation Data

- Speed
- Count (with 15%; R^2; Slope)
 - 15 minute periods
 - Targets:
 - 80% counts within 15%
 - R^2 & slope between 0.92 & 1.08





Speed Contours

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Real Data VS

Predicted Data



May 18th, 2015



Related Projects

- SANDAG DTA-ABM Integration
- USDOT ATDM-DMA Testbed
- Post Deployment AMS Evaluation
- MIT Optimization Project
- Future ICM Corridors (13 USDOT Grants)



Why DTA-ABM Integration

- ABM Temporal Integration
 - Choice models
- Congestion Duration
- Dynamic Tolling
- Travel Time Reliability
- Refined Speeds for Air Quality Analysis
- Launching Point for New ICM Corridors





What DTA Brings to RTP

- Operational Analysis
 - ITS/System management applications
 - Signal timing
 - Transit improvements/projects
 - Connected Vehicles







USDOT AMS ATDM-DMA

- ATDM Strategies
 - Network Prediction Sensitivity
 - Dynamic routing
 - Dynamic Pricing
 - Dynamic Lane Use Control
 - Dynamic Managed Use Lanes





- DMA Strategies
 - INFLO
 - Queue Warning (Q-Warn)
 - Dynamic Speed Harmonization (SPD-HARM)
 - Cooperative Adaptive Cruise Control (CACC)
 - MMITSS
 - Intelligent Traffic Signal System (ISIG)
 - Transit Signal Priority (TSP)





Questions

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