



Travel Time Prediction Using Higher Traffic Flow Models

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Outline of the presentation

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Introduction

- Travel time is an important system performance measure.
- Availability of travel time information
 - Control traffic and reduce congestion.
 - Identification of shortest path between a pair of origin and destination.
 - Essential for various Intelligent Transportation Systems (ITS) applications such as Advanced Traveler Information Systems (ATIS), Advanced Public Transportation System (APTS) and Advanced Traffic Management Systems (ATMS).
 - Dynamic route guidance, incident detection, freeway ramp metering control etc.
- Travel time - a spatial parameter- Prediction is challenging.

Background



- Various techniques - reported in literature.
 - Data driven approach
 - Traffic flow theory based approach
- Data driven – extracts system characteristics from the huge amount of data – data intensive.
- Traffic flow theory based approach – concepts of physics – relates traffic flow variables – complete picture of the system- aggregate level- minimal data.
- Developing countries- Minimal data collection infrastructure.
- Travel time prediction using traffic flow theory based approach – Suitable.



Back ground

- Traffic flow models – Modified/Rewritten/discretized/state space – travel time prediction/estimation.
- Macroscopic traffic models – First order models- Higher order models.
 - Models: LWR, Payne, Zhang, Aw-Rascle.
- First order
 - LWR- simplest – explored to an extent.
- Higher order models
 - Payne- Negative speeds – Characteristics speed higher than vehicle speed.
 - Aw-Rascle– Answers the limitations of Payne’s model -not yet explored.
- Adopted – Aw-Rascle’s traffic flow model– Travel time prediction.



Aw Rascle Model

$$\frac{\partial \rho}{\partial t} + \frac{\partial(\rho v)}{\partial x} = 0 \quad (\text{Conservation equation})$$

$$\frac{\partial(v + p(\rho))}{\partial t} + v \frac{\partial(v + p(\rho))}{\partial x} = \frac{1}{T}(u_e(k) - u) \quad (\text{Velocity dynamics equation})$$

Where,

v is speed,

ρ is traffic density,

p is the traffic pressure and

t is the independent variables represent time.

x is the independent variables represent space.

Modelling



- Purely hyperbolic and conservative form of Aw-Rascle Model i.e. under no diffusion and no relaxation conditions (Aw & Rascle, 2000).

$$\frac{\partial \rho}{\partial t} + \frac{\partial(\rho v)}{\partial x} = 0 \quad (1)$$

$$\frac{\partial(v + p(\rho))}{\partial t} + v \frac{\partial(v + p(\rho))}{\partial x} = 0 \quad (2)$$

- Taking the velocity dynamics equation and modified for travel time prediction.
- Assumption
 - Equations are independent to each other

Modelling

- Adopting the pressure function suggested by Zhang et al., (2016)

$$p(k) = v_f - V(k) \quad (3)$$

- Assuming Greenshield speed-density function for $V(k)$, eqn. (3) becomes

$$p(k) = v_f \left(\frac{k}{k_j} \right) \quad (4)$$

- Subs (4) in (2), the modified form of Aw-Rascle's velocity dynamics equation is

$$\frac{\partial \left(v + v_f \left(\frac{k}{k_j} \right) \right)}{\partial t} + v \frac{\partial \left(v + v_f \left(\frac{k}{k_j} \right) \right)}{\partial x} = 0 \quad (5)$$

Modelling



- Discretization: Finite difference method
 - Forward Time Backward Space scheme

$$\frac{\partial u}{\partial x} = \frac{u_i^t - u_{i-1}^t}{\Delta x} \quad \frac{\partial k}{\partial x} = \frac{k_i^t - k_{i-1}^t}{\Delta x}$$

$$\frac{\partial u}{\partial t} = \frac{u_i^{t+1} - u_i^t}{\Delta t} \quad \frac{\partial k}{\partial t} = \frac{k_i^{t+1} - k_i^t}{\Delta t}$$

Where,
 t is the time index
 i is the space index

- Discretizing the eqn. (5) and rewriting as (6)

$$u_i^{t+1} = u_i^t - \left(\left(\frac{v_f k_i^{t+1}}{k_j} \right) - \left(\frac{v_f k_i^t}{k_j} \right) \right) - \left(\frac{\Delta t}{\Delta x} \right) u_i^t \left[(u_i^t - u_{i-1}^t) + \left(\left(\frac{v_f k_i^t}{k_j} \right) - \left(\frac{v_f k_{i-1}^t}{k_j} \right) \right) \right]$$
$$u_i^{t+1} = u_i^t - \left(\frac{v_f}{k_j} (k_i^{t+1} - k_i^t) \right) - \left(\frac{\Delta t}{\Delta x} \right) u_i^t \left[(u_i^t - u_{i-1}^t) + \left(\frac{v_f}{k_j} (k_i^t - k_{i-1}^t) \right) \right] \quad (6)$$

Implementation

- Assumptions

- First sections(Entry) speed and density values – initial conditions.
- Overtaking of vehicles are permitted
- Stability Condition of the numerical scheme: $\left(\frac{\Delta x}{\Delta t}\right) > v_f$
(vehicle travelling at V_f cannot travel more than one time step)

- Adopting the discretized form of conservation eqn. suggested by Kumar et al., (2011).

$$k_i^{t+1} = k_i^t \left(1 - \left(\frac{\Delta t}{\Delta x}\right) v_f \left(1 - \frac{k_i^t}{k_j} \right) \right) - k_{i-1}^t \left(\left(\frac{\Delta t}{\Delta x}\right) v_f \left(1 - \frac{k_{i-1}^t}{k_j} \right) \right)$$

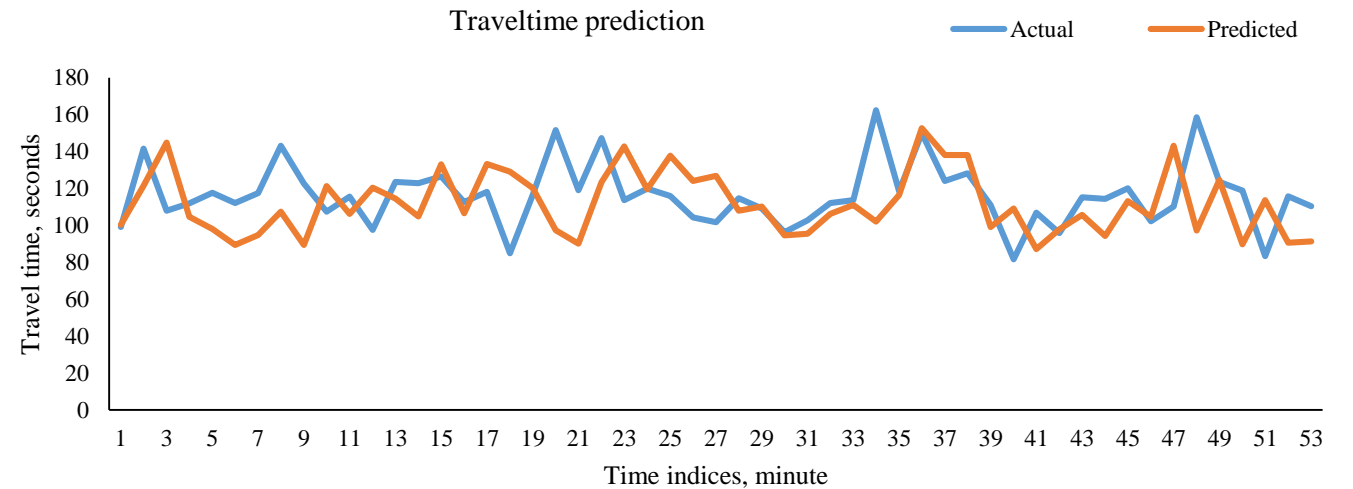
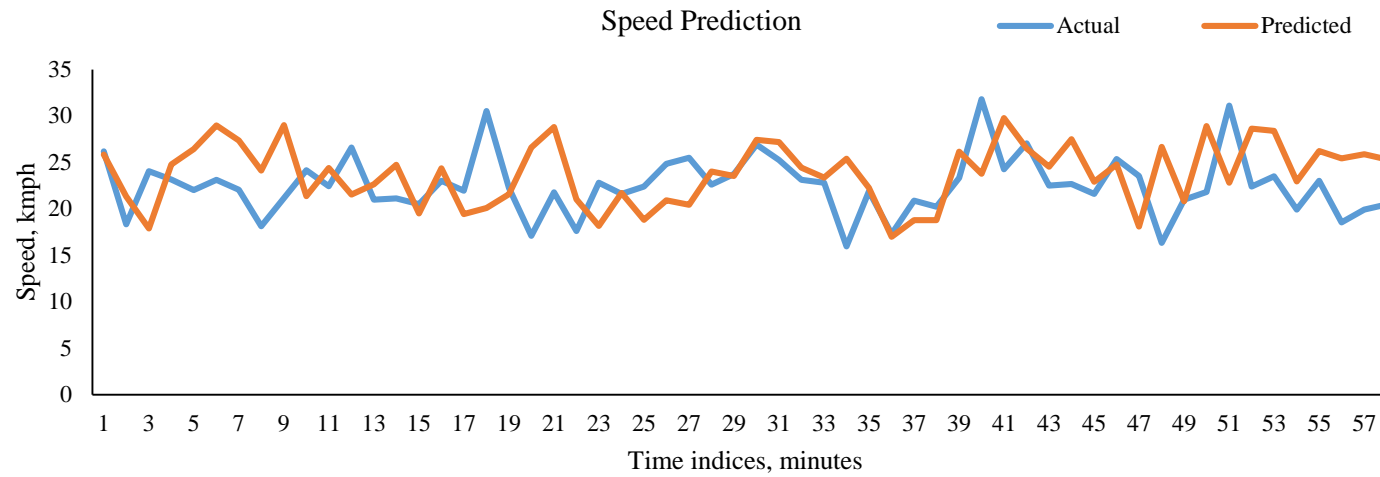
Implementation

- Study stretch : Rajiv Gandhi Salai (IT Corridor), Chennai, India.
 - Section 1: Location A to Location B
 - Section 2 : Location B to Location C
 - Total Length: 1.72 km
 - Six lane highway road
 - Only direction of traffic considered
- Road stretch similar to the field test bed was created in VISSIM using a satellite image
- Five hours simulation – implementation
- Density and speed values for both section 1 and 2 - Collected



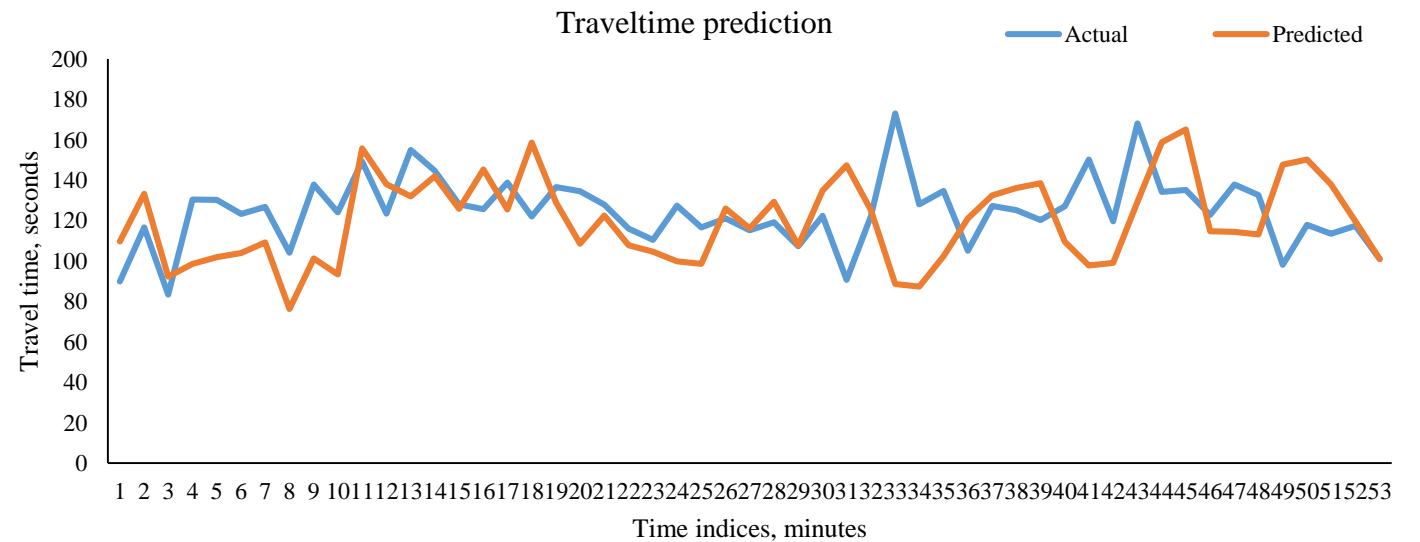
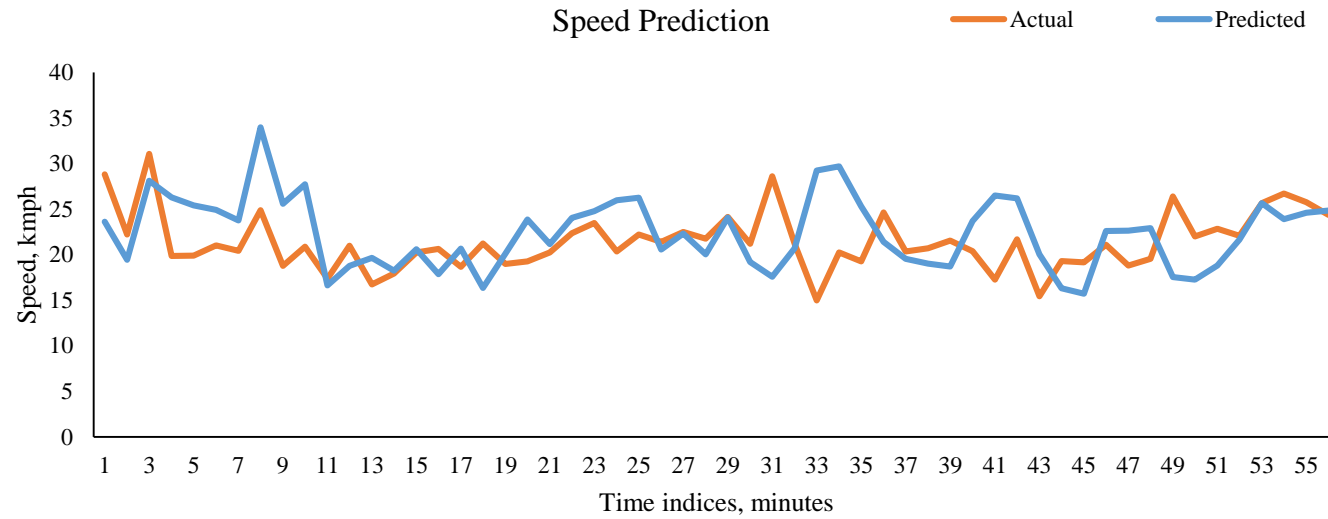
Figure: Test bed

Results – Sample 1



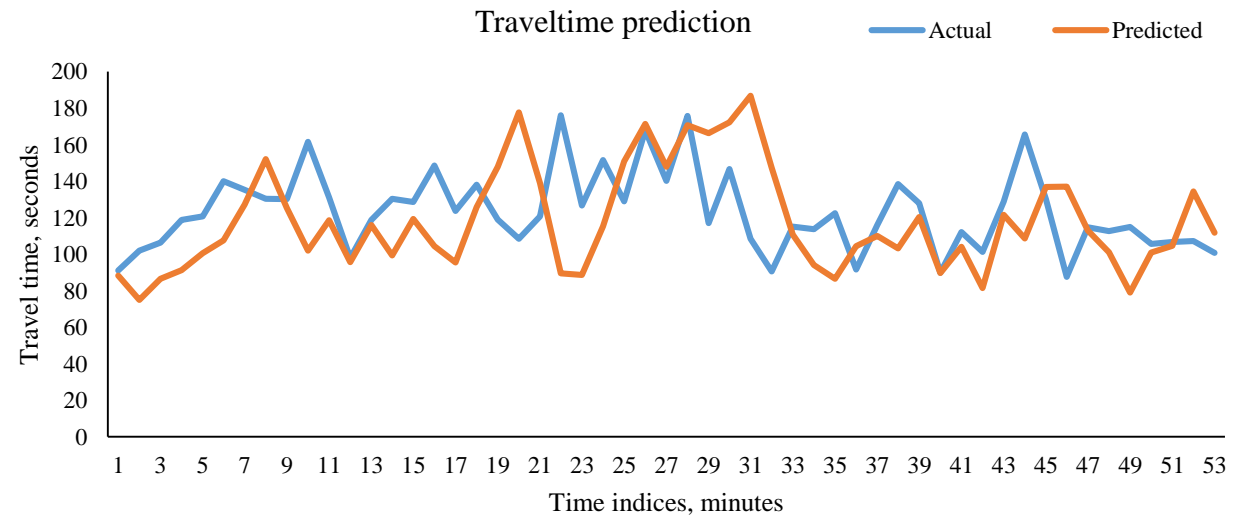
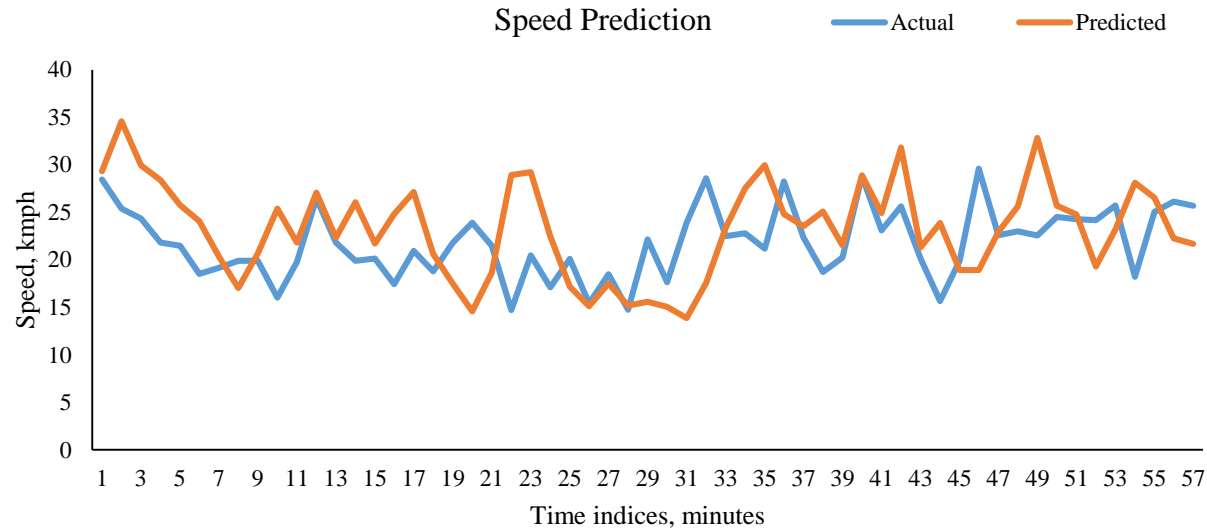


Results – Sample 2





Results – Sample 3





Conclusions

- Explored the possibility of travel time prediction using higher order traffic flow model equations.
- MAPE-15% to 21% - Satisfactory
- Model is very sensitive towards the input – variations in output - higher errors in certain places
- Initial implementation - need refinements – better predictions



Future scope

- Exponential speed density relationship.
- State space representation integrated with filtering algorithm for real time prediction.
- Possibility of using another source of data in the correction step.
- Finite Volume discretization

Thank you.