

# Traffic Prediction in Urban Road Networks

## - Comparison of Regressive Approach & Pattern Approach



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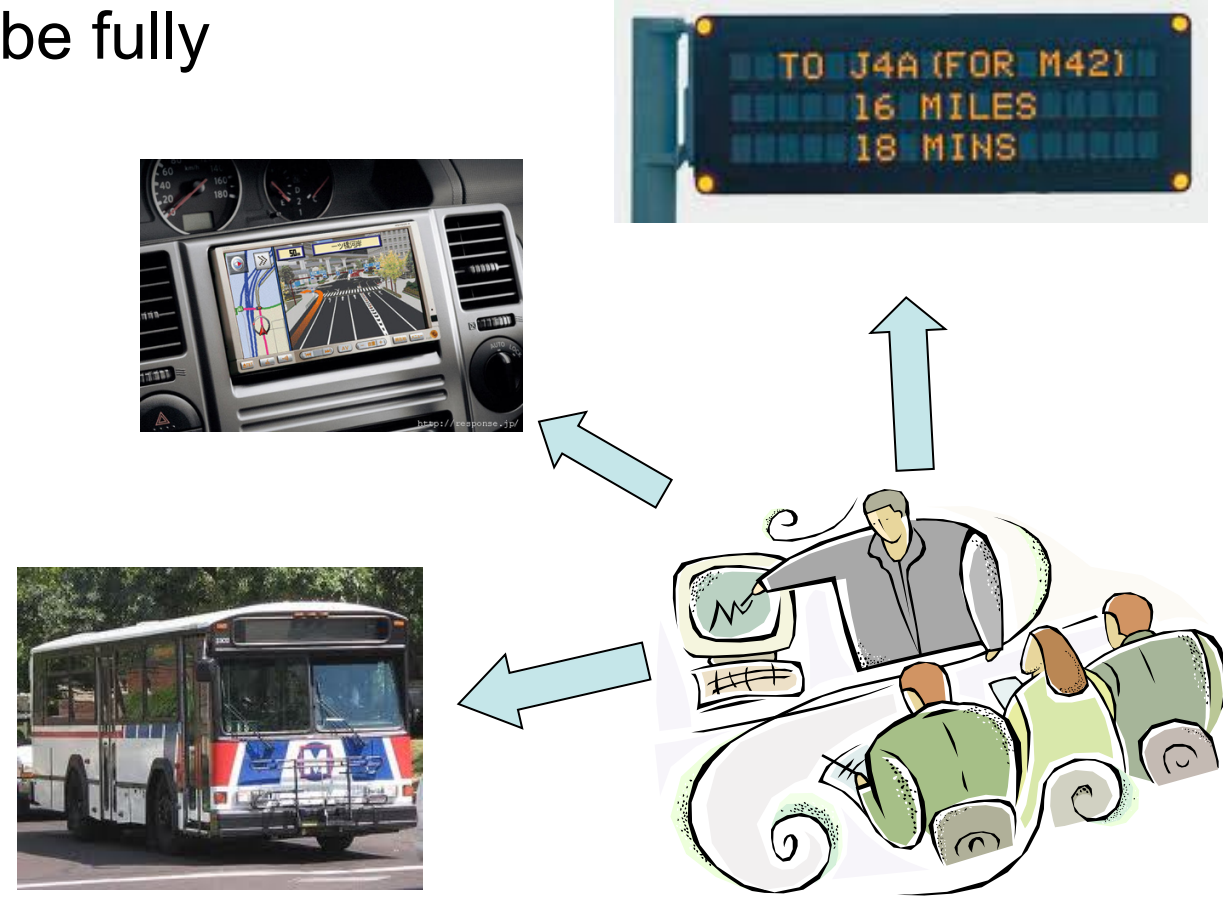
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### Introduction

• Intelligent Transportation System (ITS) refers to information and communication technology that improves transportation outcome.

• **However**, the benefits of ITS cannot be fully realized if traffic is not “known” in advance or forecasted.



### Related Work: Previous Efforts

#### • Regressive / Statistical Typical Approaches

- Auto-regressive integrated moving average (ARIMA)
- Exponential Smoothing (ES)
- Neural Network Models (NNet)
- Non-parametric Regression (NPR)

Limitation: lack of domain knowledge (e.g., people travelling behavior)

#### • Other Approaches

- Historical Average Models

Limitation: no way to react to dynamic changes, such as events

### Forecasting Problem & Approaches

#### • Problem Formulation

- Input: speed time series (i.e.,  $\{y_i\} \ 1 \leq i \leq N$ )
- Output: future speed values in M time stamps.  
i.e.,  $\{y_k\} \ (N+1 \leq k \leq N+M)$

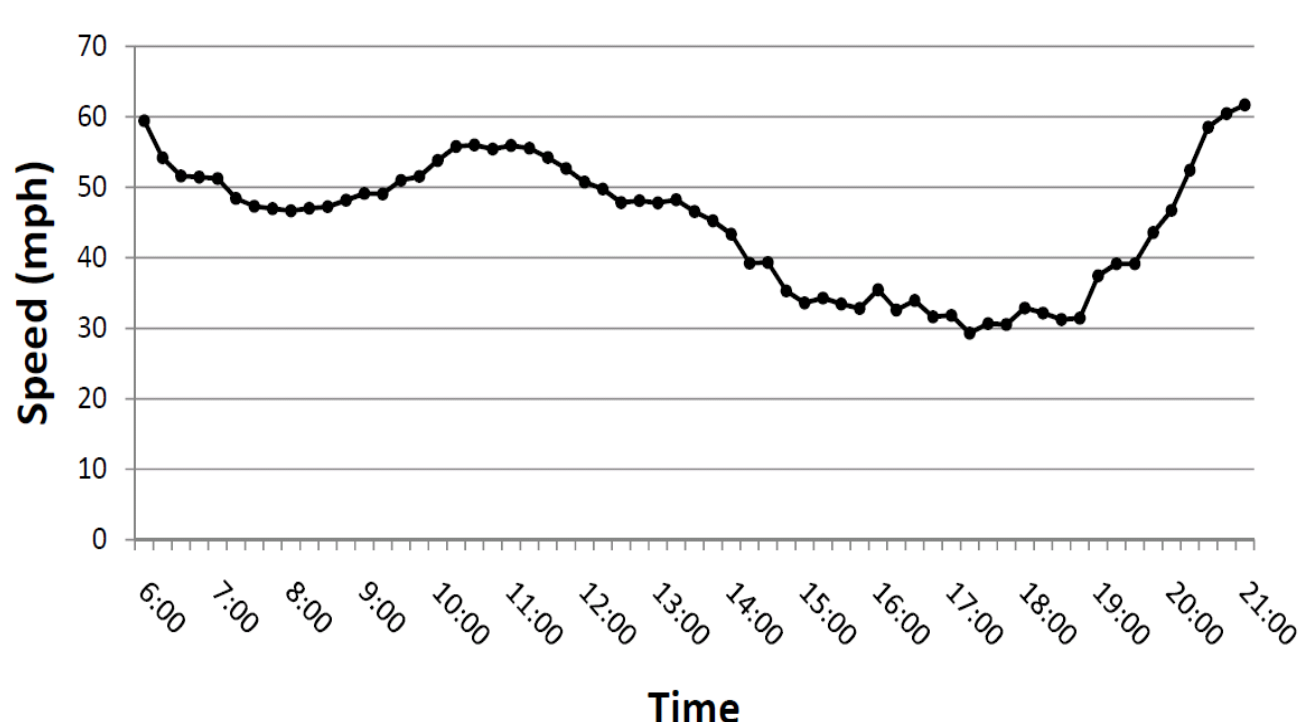
#### • Regressive Solution

- ARIMA Model
- Key Features:
  - Models the relationship between current variable with **previous variables & noise**.
  - Only predicts 1 variable ahead, and accuracy may decrease if use predicted values as known values

$$y_t = \sum_{k=1}^p \phi_k \cdot y_{t-k} + \sum_{k=1}^q \theta_k \cdot \varepsilon_{t-k} + \varepsilon_t$$

#### • Pattern Solution

- Pattern: daily speed sequence (itself is used as predicted value).
- Key Features:
  - Patterns are integrated with **people traveling behavior**, e.g., people go to work Mon-Fri, do not go to work at weekends.
  - Patterns filter out all the noises, so it is incapable of predict events.



Sample Pattern Visualization

- Pattern Categories:
  - Regular: Mon~Sun
  - **Long Weekend**: Fri~Tue
  - **Thanks Giving**: Wed~Mon
  - **Special Event**: Sat, Sun

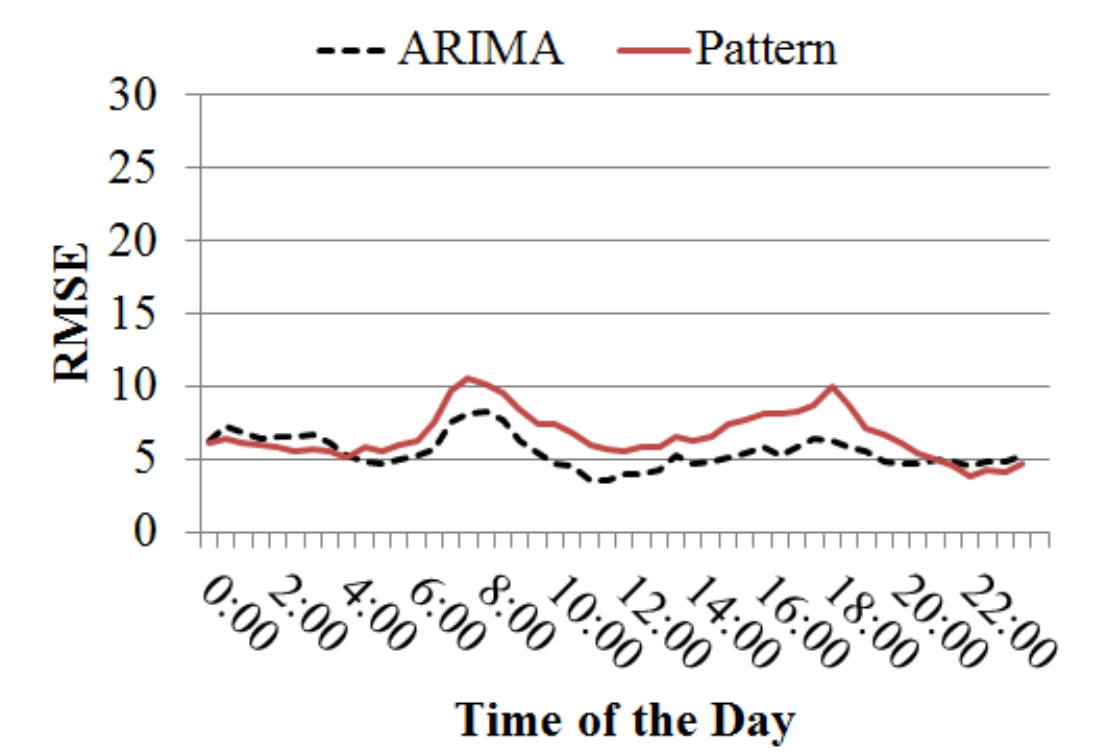
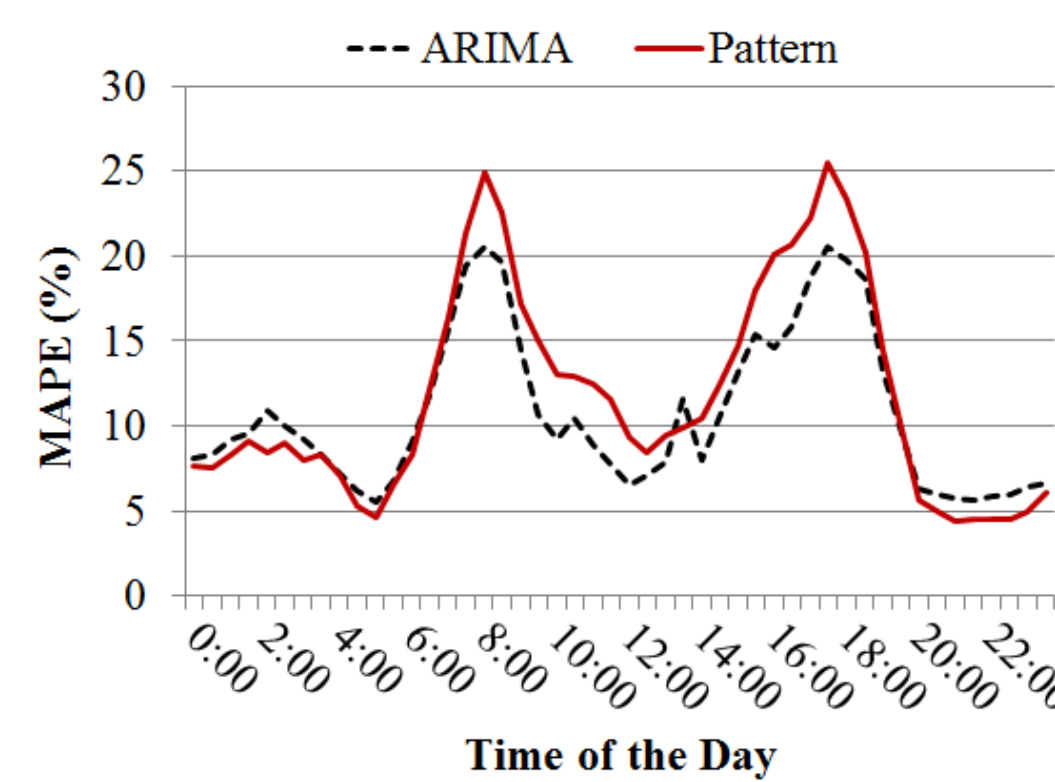
### Experiments

#### • Experimental Setup

- Spatial Coverage: 450 sensors from Los Angeles freeway
- Training data: Nov, 2011 / Testing data: first week in Dec, 2011
- Data Resolution: 5-min
- Forecasting interval: 30 minutes (i.e.  $M = 30 / 5 = 6$  time stamps)
- **Measures of effectiveness**

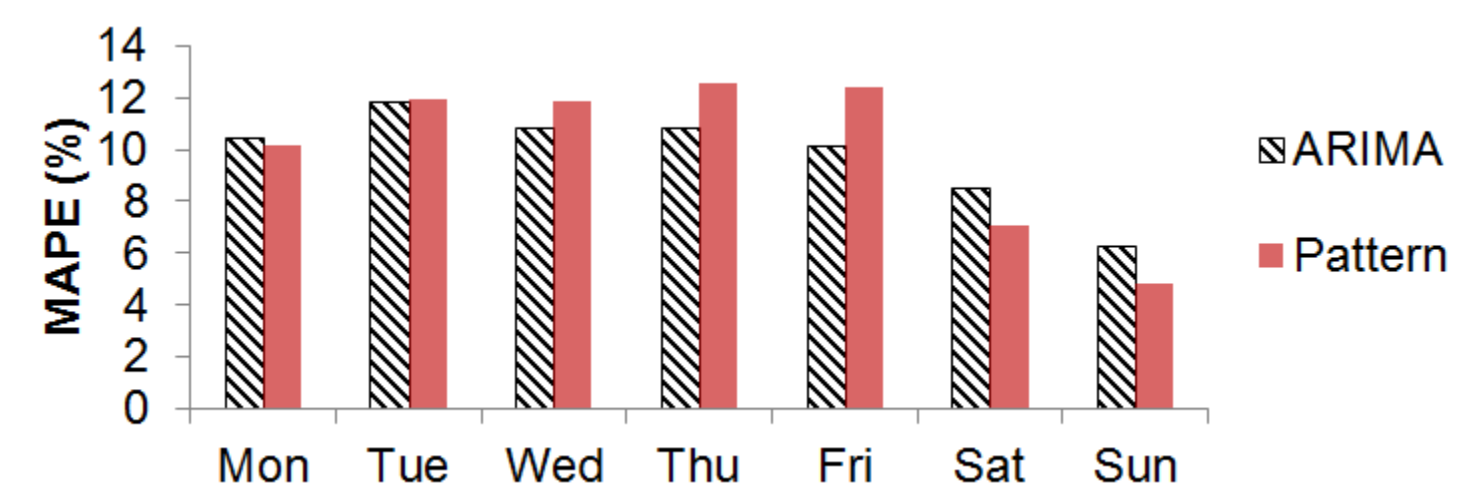
$$\text{MAPE (\%)} = \frac{1}{n} \sum_{i=1}^n \frac{|y_i - \hat{y}_i|}{y_i} \times 100 \quad \text{RMSE} = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

#### • Effects of Hours in a Day (All Sensors, Wednesday)



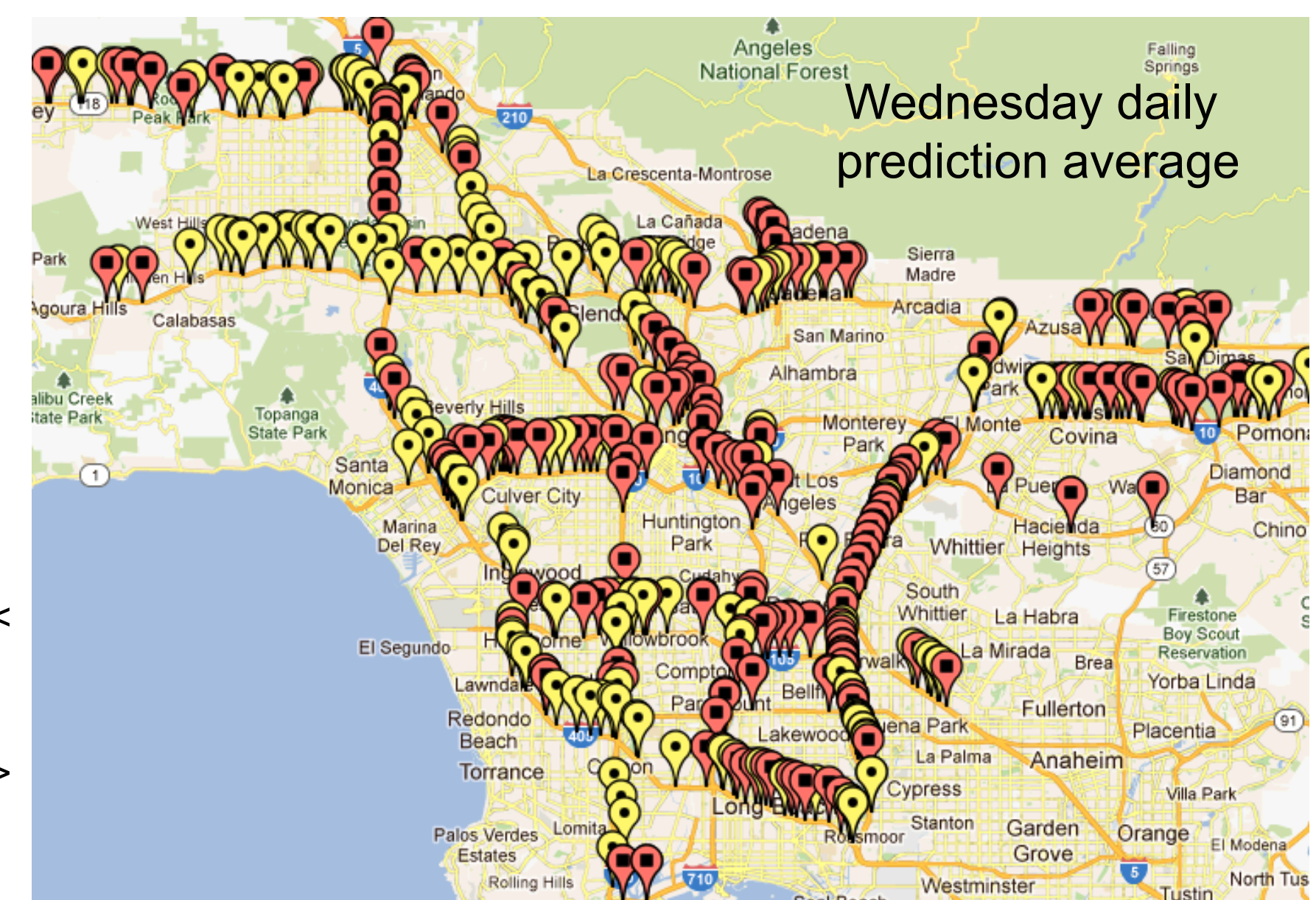
- Observations:
  - 1) Rush hour congestion makes prediction harder
  - 2) Pattern approach is better when there is free flow traffic

#### • Effects of Days in a Week (All Sensors, All hours)



- Observations:
  - 1) ARIMA approach provides better prediction during weekdays, because weekday data takes a larger portion of the training data.
  - 2) Pattern approach is better when it comes to weekend

#### • Effects of Location



- Observations:
  - 1) ARIMA approach provides better prediction for specific regions or freeways, for example, downtown area, I-10, I-605.
  - 2) Pattern approach is better when it comes to less-event area, where traffic mostly follows the cyclical nature.

### Conclusion & Future Work

#### • Conclusion

- This study empirically compares the pros & cons of regressive approach and pattern approach on traffic prediction problem.

#### • Future work

- Introduce domain knowledge while training the regressive model.
- Enrich the pattern categories with event/accident pattern etc.