

Traffic Profile Prediction

Collection, Cleansing, and Analysis of Traffic, Event, and Weather Data

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NATMEC 2016 Innovations in Traffic Data Collection and Processing





- Types of Data
- Data Collection
- Data Cleansing
- Data Analysis



Types of Data

Occupancy

- Roadway data
 - Speed
 - Volume
 - Occupancy
 - Calculated Length
- Event data
 - Incidents from the traffic database
 - Sports Events
- Condition data
 - Weather
 - Holidays
 - Calendar









Traffic data

- Raw data files
 - One file per day
 - All detectors from the district
 - 30 second or 5 minute data
- Averaged data
 - Database table
 - 15 minute rolled up data
 - Selectable by lane, detector, etc.

TSS-100620141.dat							
1	timestamp,	detector_id, lane_id,	speed, volume, occupan	cy,	clas	s_bi	in1,
2	11:58:57,	3081:tss:detector:MDX,	1322230:tss:lane:MDX,	55,	10,	20,	2,
3	11:58:57,	3081:tss:detector:MDX,	1322231:tss:lane:MDX,	58,	10,	20,	2,
4	11:58:57,	3081:tss:detector:MDX,	1322232:tss:lane:MDX,	69,	10,	20,	2,
5	11:58:57,	3081:tss:detector:MDX,	1322233:tss:lane:MDX,	67,	10,	20,	2,
б	11:59:27,	3064:tss:detector:MDX,	1322199:tss:lane:MDX,	67,	10,	20,	2,
7	11:59:27,	3064:tss:detector:MDX,	1322200:tss:lane:MDX,	61,	10,	20,	2,
8	11:59:27,	3064:tss:detector:MDX,	1322202:tss:lane:MDX,	66,	10,	20,	2,
9	11:59:27,	3064:tss:detector:MDX,	1322203:tss:lane:MDX,	56,	10,	20,	2,
10	11:59:25,	3061:tss:detector:MDX,	1322204:tss:lane:MDX,	64,	10,	20,	2,
11	11:59:25,	3061:tss:detector:MDX,	1322205:tss:lane:MDX,	59,	10,	20,	2,
12	11:59:25,	3061:tss:detector:MDX,	1322206:tss:lane:MDX,	51,	10,	20,	2,
13	11:59:25,	3061:tss:detector:MDX,	1322207: <u>tss</u> :lane:MDX,	54,	10,	20,	2,





Focused analysis on I-4 Corridor FDOT D5





SunGuide Database
Detector Data
Event Data

- Accidents
- Lane Closure
- etc.



Event data

- Final state or archive of transitions available
- Data not consistently entered
- Mix of causal and resulting data
- Weather data
 - Mix of formats available via APIs
 - Less data for historical weather
 - Most consistent is Airport data
 - Only approximately hourly
 - Not all airports are open 24 hours
 - Many variables available
 - Not all variables available all the time







- External Data
- Generated or scraped
- Calendar
 - Day of the week
 - Day of the year
 - Moon phase
 - Holiday
 - School in session
- Sporting Events
 - From a nearby venue



(Wrong Calendar?)





Data Cleansing

- Missing data
- Transitions from 30 second to 5 minute data
- Inconsistent calculated length



Detector configuration change (Volume Data)



Calculated Average Vehicle Length for Single Detector



 $\left(Speed * Occupancy * \frac{5280}{3600} * \Delta Time \right)$

Volume *100



Data Cleansing Calculated Lengths

- Great disparities
- Consistent for some lanes





Calculated vehicle length over time (one detector)

Short and Long Vehicle Length Outliers



Analytics - Visualizations

- Speed-Volume-Occupancy curves
- Speed-time-distance with events









Figure 2 - 24 hours of traffic speed along a segment of I-4 with Event locations and times marked

Speed-Volume-Occupancy-Throughput



Analytics - ARIMA

 AutoRegressive Integrated Moving-Average (ARIMA) model

- Very good steady-state predictions
- Lags at transitions







Analytics – ARIMA Model Selection

- Evaluated models generated from several sensors on different days
- Evaluated their root mean square error on different sensors and days
- Selected the ARIMA (0,1,4) model as most general

Evaluation of ARIMA Models									
model	2784	2835	3469	3678	3920	4353	5454	5579	avg.
Arima(1,1,1)	3.435	4.787	5.445	6.661	3.799	3.959	5.515	3.263	4.608
Arima(0,1,0)	4.619	6.315	7.438	9.025	2.5	3.316	7.419	4.358	5.624
Arima(0,1,4)	3.465	4.804	5.529	6.749	3.513	3.718	5.593	3.297	4.584
Arima(5,1,0)	3.647	5.026	5.812	7.105	3.263	3.722	5.907	3.472	4.744



Analytics - PAM

Partitioning Around Medoids

- Distance-based clustering of curves
- Medoid is the curve closest to the "center" of the cluster
- The fit of the cluster is measured using the Silhouette







Unperturbed Day Generation

- Used PAM to find most frequent cluster
- Used the curves within 5-10 percent of the distance of the medoid to generate unperturbed day
- Threw out most extreme day for each 15-minute interval
- Used the average of the remaining points to define the unperturbed day







Analytics - Events

- Events are not at detectors
- Event start and stop times are not well recorded
- Events may have impacts upstream and downstream of the event itself







Analytics - Weather

- Weather has measurable impacts on traffic predictions
- Real-time and future weather data may be more granular than historical data







- Data from a wide variety of sources is needed to predict traffic profiles
- Data for predicting traffic profiles frequently needs to be cleansed
- Algorithms for predicting traffic profiles must be able to deal with missing data
- The algorithm for predicting an unperturbed day provides a baseline for finding events
- Events, conditions, and weather have impacts on traffic profiles that must be taken into account to get good predictions



Technologies Used

Spark 152

Workers

Spark Master at spark://

URL: spark: REST URL: spark (cluster mode) Alive Workers: 2 Cores in use: 16 Total, 0 Used Memory in use: 28.9 GB Total, 0.0 B Used Applications: 0 Running, 3 Completed Drivers: 0 Running, 0 Completed Status: ALIVE







 Worker Id
 Address
 State
 Cores
 Memory

 worker-20160325200439 ALIVE
 8 (0 Used)
 14.4 GB (0.0 B Used)

 worker-20160325200605 ALIVE
 8 (0 Used)
 14.4 GB (0.0 B Used)

Running Applications

Application ID	Name	Cores	Memory per Node	Submitted Time	User	State	Duration

Completed Applications

Application ID	Name	Cores	Memory per Node	Submitted Time	User	State	Duration
app-20160327145138- 0002	Create Rollup	24	1024.0 MB	2016/03/27 14:51:38	dataminer	FINISHED	33 min
app-20160325235534- 0001	CSV -> Dataframe (parquet)	24	1024.0 MB	2016/03/25 23:55:34	dataminer	FINISHED	33 min
app-20160325200822- 0000	CSV -> Dataframe (parquet)	24	1024.0 MB	2016/03/25 20:08:22	dataminer	FINISHED	33 min



Hadoop Overview Datanodes Snapshot Startup Progress



Shedoop

Security is off. Safemode is off.

2876 files and directories, 5575 blocks = 8451 total filesystem object(s).

Heap Memory used 70.78 MB of 111.5 MB Heap Memory. Max Heap Memory is 889 MB.

Non Heap Memory used 42.12 MB of 61.69 MB Committed Non Heap Memory. Max Non Heap Memory is 214 MB

Configured Capacity:	14.44 TB
DFS Used:	774.63 GB
Non DFS Used:	1.71 TB
DFS Remaining:	11.98 TB
DFS Used%:	5.24%
DFS Remaining%:	82.94%
Block Pool Used:	774.63 GB
Block Pool Used%:	5.24%
DataNodes usages% (Min/Median/Max/stdDev):	3.33% / 6.38% / 7.12% / 1.57%
Live Nodes	4 (Decommissioned: 0)
Dead Nodes	0 (Decommissioned: 0)
Decommissioning Nodes	0
Number of Under-Replicated Blocks	0
Number of Blocks Pending Deletion	0
Block Deletion Start Time	2/9/2016, 2:14:14 PM









Questions?

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