Data Mining in Vehicular Sensor Networks: Technical and Marketing Challenges

Hillol Kargupta

Agnik & University of Maryland, Baltimore County

http://www.cs.umbc.edu/~hillol http://www.agnik.com

Roadmap

- Motivation
- Mining vehicular sensor network
- Building MineFleet
- Challenges
- Some Algorithmic Solutions
- Discussion

Vehicles: Source of High Volume Data Streams



- Vehicles generate tons of data
- Hundreds of different parameters from different subsystems
- High throughput data streams

So what?



High gas prices

Why Mine Vehicle Data?

- Fuel consumption analysis
- Fleet analytics
- Vehicle benchmarking
- Predictive health-monitoring
- Driver behavior analytics





Bad driving costs money---fuel, brake shoe, insurance, law-suits

SPEED

Fuel Subsystem: Sample Attributes

Fuel Subsystem

- Air Fuel Ratio
- Fuel Level Sensor (%)
- Fuel System Status Bank 1 [Categ. Attrib.]
- Oxygen Sensor Bank 1 Sensor 1 [mV]
- Oxygen Sensor Bank 1 Sensor 2 [mV]
- Oxygen Sensor Bank 2 Sensor 1 [mV]
- Oxygen Sensor Bank 2 Sensor 2 [mV]
- Long Term Fuel Trim Bank 1 [%]
- Short Term Fuel Trim Bank 1[%]
- Idle Air Control Motor Position
- Injector Pulse Width #1 (msec)
- Manifold Absolute Pressure (Hg)

Operating Condition

- Barometric Pressure
- Calculated Engine Load(%)
- Engine Coolant Temperature (°F)
- Engine Speed (RPM)
- Engine Torque
- Intake Air Temperature (IAT) (°F)
- Mass Air Flow Sensor 1(MAF) (lbs/min)
- Start Up Engine Coolant Temp. (°F)
- Start Up Intake Air Temperature (°F)
- Throttle Position Sensor (%)
- Throttle Position Sensor (degree)
- Vehicle Speed (Miles/Hour)
- Odometer (Miles)

Product Concept: MineFleet

Optimize Fuel Economy by

- Modeling fuel consumption behavior
- Identifying factors that are causing poor fuel economy
- Benchmarking fuel sub-system

Predictive Health Monitoring

- Automatically execute built-in library of tests for checking the health-status of the vehicle
- Predictive modeling of the vehicle sub-systems

Driver Behavior Monitoring

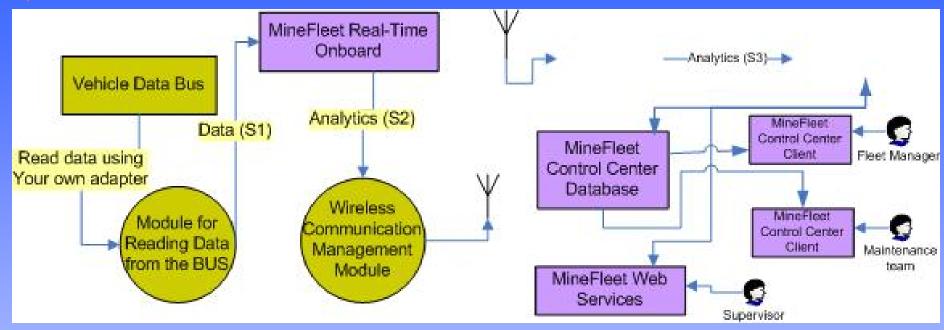
- Policy-based driver behavior monitoring
- Identify the effect of driver behavior on fuel economy

Minimize Wireless Communication

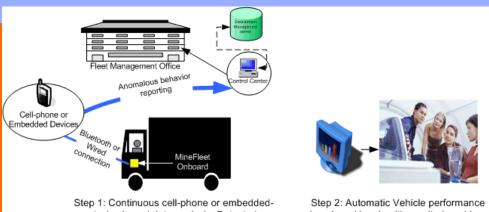
- Onboard data stream mining
- Send alerts and analytics only when problems occur

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MineFleet Architecture

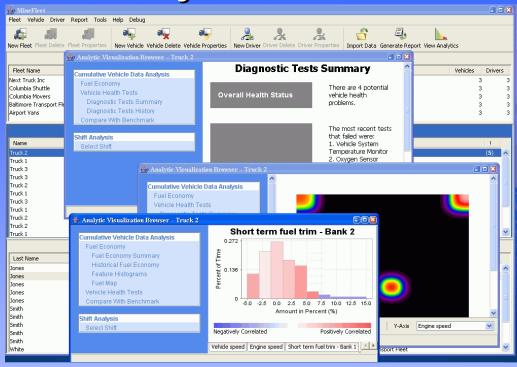


- Main Components
 - Onboard Module
 - MineFleet Control Center Server
 - MineFleet Client Modules
 - MineFleet Web Services



Step 1: Continuous cell-phone or embeddeddevice based data analysis. Detected anomalous behavior reported over wireless link. No expensive continuous data downloading over wireless network. Step 2: Automatic Vehicle performance benchmarking, health monitoring, driver behavior monitoring, fuel consumption analysis, data stream management, and report generation.

MineFleet System



MineFleet Control Center

MineFleet Onboard

Challenges: Accessing Data

- Vehicles generate data for hundreds of attributes
- But manufacturers provide open access to only about
 20 of those that are needed for emission checks
- Off-the-shelf devices were designed for off-line monitoring by mechanics

Onboard Computing Platform



- First prototype -- PDA-based platform
- Other choices:
 - Cell phones and
 - Low-cost, less powerful embedded devices

Circa 2005



Circa 2007



- Market Entry Point
 - Location management companies
 - M2M companies
- Resource constrained
- → 3-4K run time memory
- 250K footprint
- → Resource sharing with GPS program.

Fuel Economy: Impact of Vehicle Condition and **Driver Behavior**

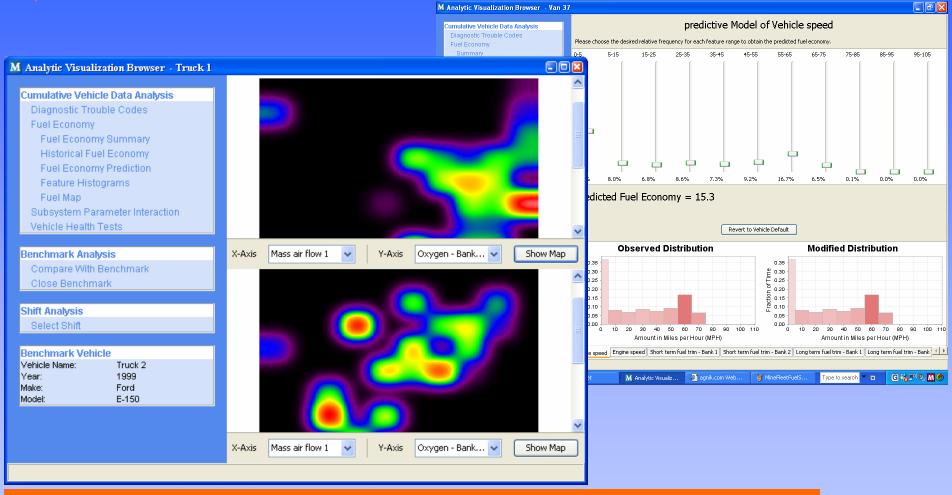
- Quantify the effect of vehicle condition on fuel consumption. Example:
 - Effect of air-intake subsystem behavior on fuel economy
 - Effect of fuel subsystem on fuel economy.
- Quantify the effect of driver behavior on fuel consumption.
 - Effect of speeding on fuel economy
 - Effect of acceleration on fuel economy
 - Effect of braking on fuel economy
 - Effect of idling on fuel economy



Poor vehicle components and bad driving reduces gas mileage

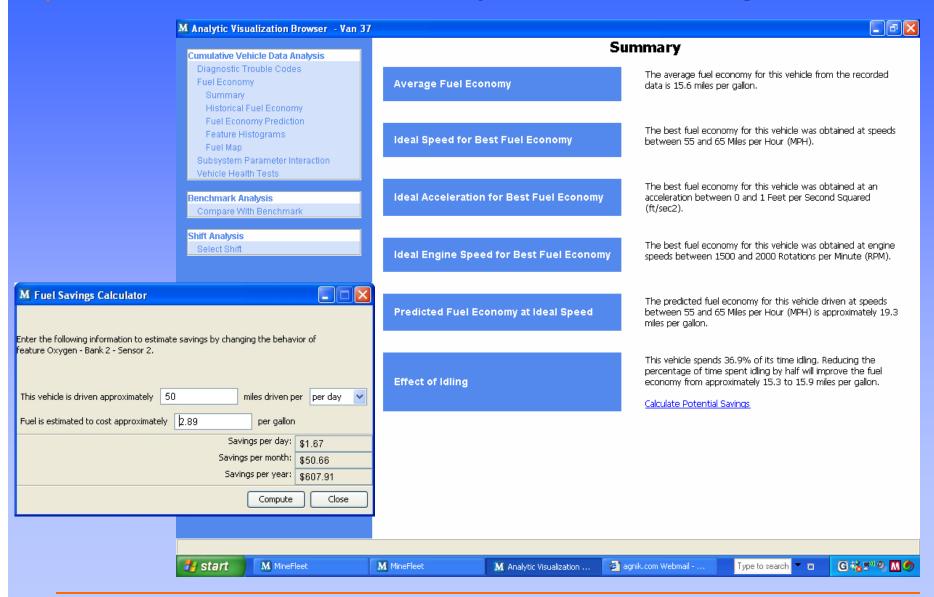
Build predictive models of the fuel economy as a function of vehicle and driving parameters for optimizing the performance

Fuel Heat Map & Predictive Modeling



Fuel heat maps show the vehicle operating points that offer high fuel economy. Red color represents high fuel economy and blue represents poor.

Fuel Consumption Summary Panel & Savings Calculator



Predictive Vehicle Health Management

Detect problems using physics-based model and inductive techniques.



Vehicle Diagnostic Tests

Air Intake Volume Inconsistency Engine Intake Vacuum Inefficiency

Air Intake Management Monitor

Thermal Event Detector Throttle Request Status

Failed Diagnostic Test

Idle Air Control

Combustion Temperature Inequality Monitor

Oxygen Sensor Operating Condition Monitor

Long Term Fuel Related Combustion Efficiency

Air Intake Volume Inconsistency

Combustion Temperature Control Decay Monitor

Long Term Fuel Related Combustion Efficiency

Diagnostic Trouble Codes

Fuel Economy

Subsystem Parameter Interaction Vehicle Health Tests

Cumulative Vehicle Data Analysis

M Analytic Visualization Browser - Truck 3

Summary

Vehicle Health Tests History

Benchmark Analysis

Compare With Benchmark

Shift Analysis

Select Shift

Fuel Economy

Vehicle Health Tests

Summary

Long Term Fuel Related Combustion Efficiency

Air Intake Volume Inconsistency

Thermal Event Detector

Quantitative Fuel Management

Monitoring, Fuel 9

Vehicle System Temperature

Monitor

Transmission Lubricating

Monitor

Driving Analysis

Fault Codes Shift Properties

As part of the combustion formula, fuel delivery is also the most adaptive process the vehicle has to long term wear during normal engine operation. While there is system failure codes associated with reaching the limitations of its adaptability, often significant collateral breakdown has occurred before the actual code will set. This test is designed to monitor changes well before they reach the breakdown stage. By monitoring these changes within the fuel delivery portion as they occur over time, we can often preempt the collateral damage through early detection of the deterioration.

Long term fuel trim out of range

Test Failed

MineFleet recommends checking fuel pressure (too high), injectors for leakage, leaking fuel pressure regulator, clogged evaporative emissions system, oxygen sensor contamination and clogged air filter as most likely causes when fuel trim falls high. MineFleet recommends checking for clogged injector(s), ignition system components, fuel pressure (low), or water intrusion on oxygen sensor as possible causes. Identifying all the vehicles in a fleet with a

Vehicle Diagnostic Tests

Deselect All

Select All

Flagged Vehicles

Name

Truck 2

Truck 4

Truck 6

Truck 1

Truck 3

Truck 5

Truck 2

Truck 4

Truck 6

Truck 1

Truck 3

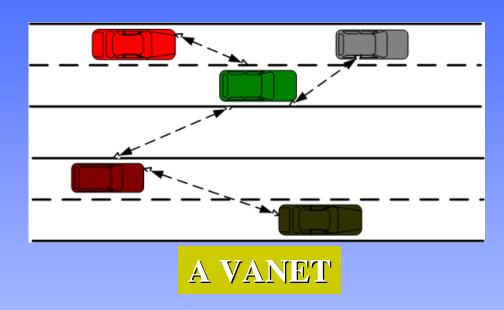
Truck 5

specific problem

Detailed description of a specific test that the vehicle passed

MineFleet VANET Project





- Developing a mobile data stream management system for quick indexing and retrieval of information from the device onboard the vehicle.
- Distributed indexing and clustering techniques

Algorithmic Challenges

- Ensemble-based Approach
- Exact vs. Approximate techniques
- Approximate monitoring of statistical properties
 - For example, Correlation Matrix
- Approximate sequence comparison
- Approximate modeling
- Similarity preservation, approximation and orthogonality
 - Fourier, Wavelet, Eigenvectors, Random vectors

Correlation Matrix Computation & Monitoring

- Given data matrix X
- Naïve computation: Compute X^TX
- Compute in the frequency domain (take Fourier transformation)
- StatStream (Zue and Shasha, 2002)
- Our Approach Exploits
 - Divide and Conquer
 - Approximate orthogonality of random vectors
- Identify the region of the matrix that contain significantly changed coefficients

Testing A Group of Correlation Coefficients Together

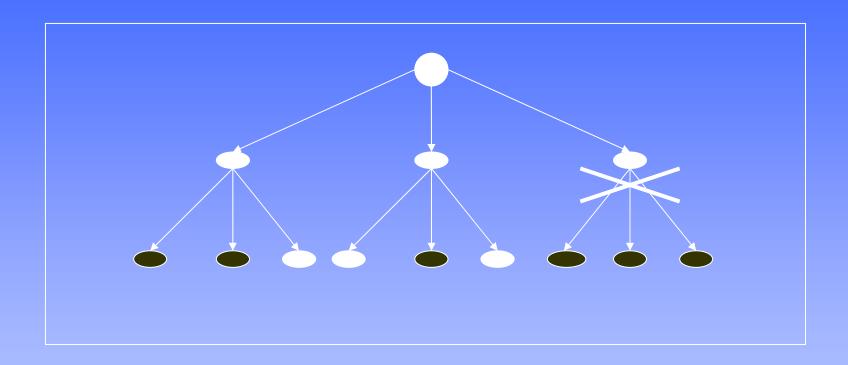
Given a subset of attributes:
$$x = \{x_1, x_2, \dots, x_k\}$$

A random vector
$$\boldsymbol{\sigma} = \{\boldsymbol{\sigma}_1, \boldsymbol{\sigma}_2, \dots, \boldsymbol{\sigma}_k\}$$

Compute
$$s = x \sigma^T$$

$$\frac{1}{r} \sum_{p=1}^{r} \text{Variance}(s)^2 \approx \sum_{l,q} \text{Correlation}(x_l, x_q)^2$$

Divide-and-Conquer Search for Significant Correlation Coefficients



- Impose a tree-structure:
 - Leaf node: a unique correlation coefficient
 - Root of a sub-tree: set of all coefficients corresponding to the leaves in that sub-tree

Variational Approximation

- Formulate as an optimization problem
- Introduce approximations
- Example: Finite Element Technique

Solve
$$-u''(x) = f(x), x \in (a,b), u(a) = u(b) = 0$$

Equivalent to minimizing
$$J(u) = \int_{a}^{b} (u^{*'}(x) - u'(x))^{2} dx$$

Continued

 Introduce approximation using locally decomposable representation

Example:
$$u(x) = \sum_{i} \alpha_{i} \gamma(x)$$

Plug-in the approximation in the objective function

Regression: Variational Formulation

Minimize
$$J(w) = \frac{1}{2}(w^* - w)^T C(w^* - w)$$

where
$$C_{j,k} = \sum_{i=1}^{n} x_{i,j} x_{i,k}$$

Can be reduced to minimizing $J(w) = -w^T b + \frac{1}{2} w^T C w^T$

$$w_{j} = \frac{b_{j} - \sum_{k \neq j} C_{j,k} w_{k}}{C_{j,j}}$$

Inner Product Computation

$$C_{j,k} = \sum_{i=1}^{n} x_{i,j} x_{i,k}$$

Approximate Inner Product Computation

- Deterministic Techniques
 - Orthogonal Transformations
- Probabilistic Techniques
 - Random vectors

Approximate Inner Product Computation

Egecioglu and Ferhatosmanoglu, 2000

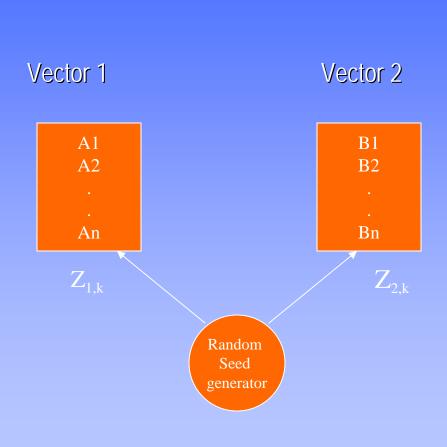
$$u.v \approx (b_1 \Psi_1(u) \Psi_1(v) + b_2 \Psi_2(u) \Psi_2(v))^{1/2}$$

$$\Psi_p(u) = \sum_{i=1}^n u^p \text{ for } p = 1,2$$

b1 and b2 can be found by minimizing the error

$$\int ((u.v)^2 - (b_1 \Psi_1(u) \Psi_1(v) + b_2 \Psi_2(u) \Psi_2(v))^{1/2}) du dv$$

Approximate Inner Product Computation



Node 1 computes Z_{1,k}

$$\Box Z_{1k} = A1.J_1 + ... + An.J_n$$

□ $J_i \in \{+1,-1\}$ with uniform probability

- Node 2 calculates Z_{2,k}
 - $\square Z_{2k} = B1.J_1 + ... + Bn.J_n$
- Compute z_{1,k}.z_{2,k} for a few times and take the average

Discussion

- Need for light-weight algorithms for real-time embedded applications
- Data intensive sensor networks may have different needs
- Distributed data stream mining

Announcement

- National Science Foundation Symposium on Next Generation Data Mining
- www.cs.umbc.edu/~hillol/NGDM07/