

On Board Monitoring of 270 Trucks: Data Analysis Case Study

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Background

- Providing drivers feedback can help
 - enhance drivers' immediate driving performance
 - induce long-term positive changes in driving behavior (Donmez et al, 2008; McGehee et al, 2007)
- Safety/fleet managers have a great influence on the safety attitudes and motivation of drivers (Newman et al, 2008; Arboleda et al., 2003)
- Initial pilot study demonstrated that feedback to commercial drivers hold promise (Hickman et al, 2009)

Study Objective

- To examine whether safe driving habits can be enhanced and risky behavior be reduced among commercial drivers with
 - Real-time feedback
 - Coaching from safety supervisors
- On-Board Monitoring System (OBMS) for Commercial Motor Vehicle safety
 - Research program sponsored by Federal Motor Carrier Safety Administration (FMCSA)
 - In support of their mission to reduce the number and severity of crashes related to large trucks

Real-Time Feedback

- **Safety/Performance Event Notification**
 - Notify drivers when a safety or performance event has been captured for coaching.
 - Safety event capture: e.g., Large lateral acceleration, Forward collision.
 - Performance event: e.g., hard brake
- **Driver State Notification**
 - E.g., Aggressive, Inattentive
- **Imminent Crash Warning**
 - Forward Collision, Lane departure



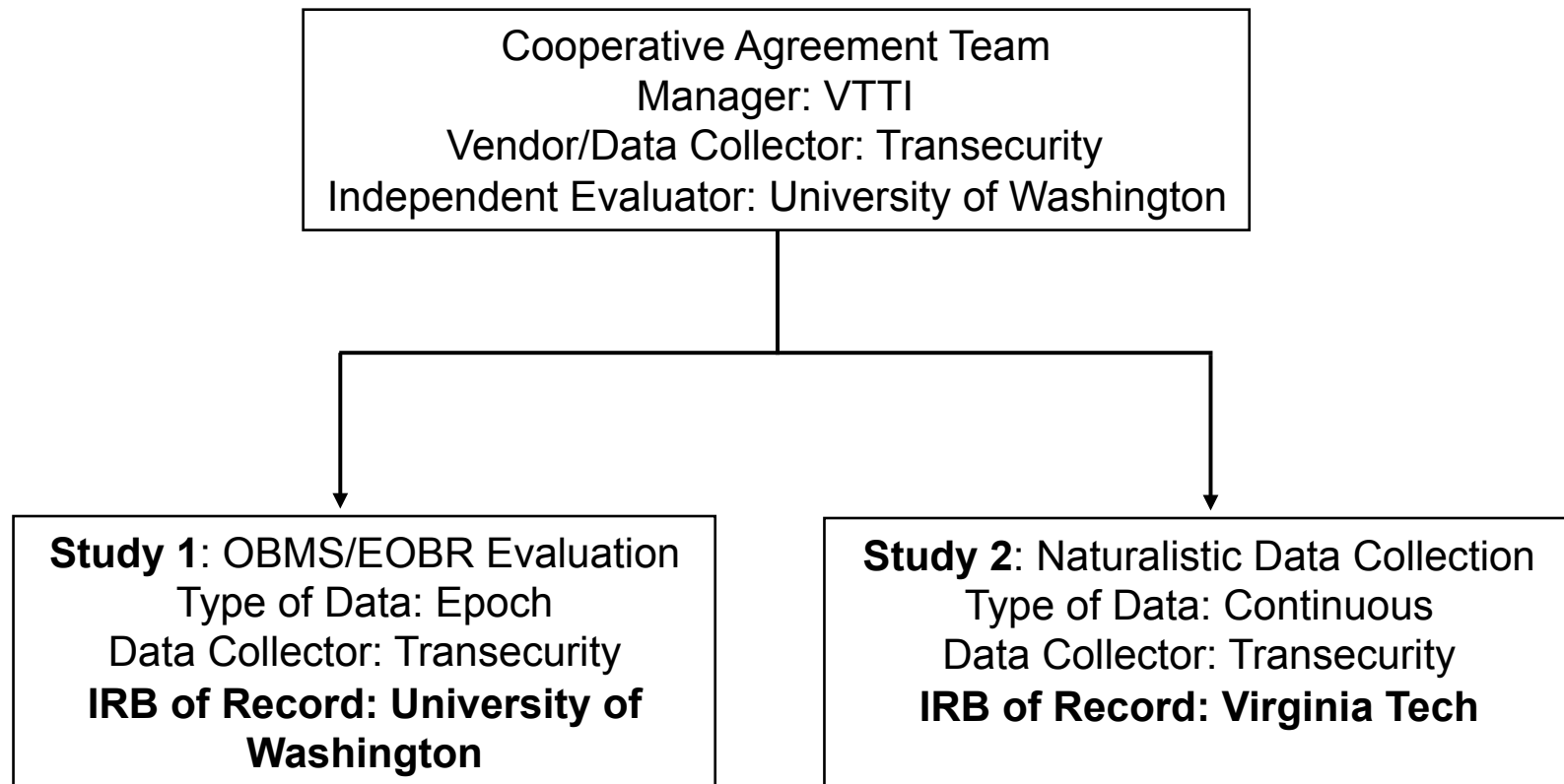
Coaching by safety supervisors

- Record snippets on unsafe driving behaviors (e.g., hard braking)
 - Video data and vehicle kinematic measures
- Reviewed by safety supervisors and used to coach the driver
- Same data will also be accessed by research team for an independent evaluation

Data collection

- 270 trucks will have OBMS devices
 - With up to 500 drivers
 - And 18 driver-months of data per driver
- Includes baseline and withdrawal period
 - Identify relative increases/decrease in performance
 - Identify any lasting or residual effects from using the system
- Includes comparison group
 - Drivers who receive no feedback for entire period
- Data will be collected on:
 - Vehicle kinematics
 - Video data
 - Driver Questionnaires

Data Collection Overview



Constraints

- Devices take time to install
 - There is a six-month installation period
 - Total data collection will be conducted over 24 months
- Maximize driver exposure to intervention method (feedback)
 - Also to account for carriers' expectation
- Need to accommodate the likely attrition of participating drivers

Experimental Design

- Three experimental groups
 - Group 1: Baseline group
 - Group 2: Longer-term adaptation group
 - Group 3: Shorter-term adaptation group

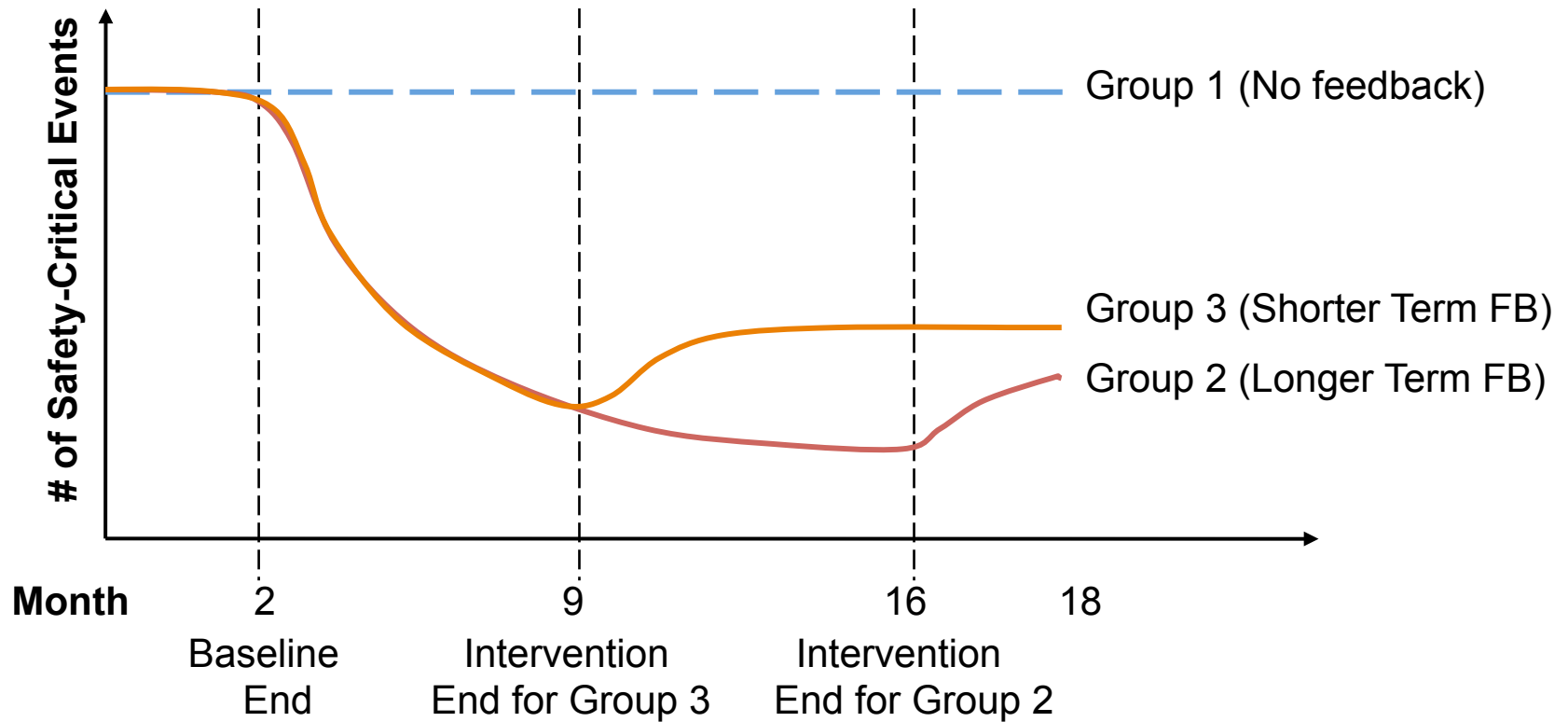
	Months																	
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18
Group 1	Baseline (no feedback)																	
Group 2	Baseline	Feedback (14 months)														WD		
Group 3	Baseline	Feedback (7 months)						Withdrawal (WD)										
	Q1	Q2	Q3					Q4								Q5	Q6	

Sample size

Group	Initial Sample Distribution
1: Baseline	30 devices
2: Longer term (14 months)	210 devices
3: Shorter term (7 months)	30 devices

- For longer term feedback (14 months), drivers that attrition out cannot be replaced given the length of feedback
 - Assuming 5 % attrition per month => 88 (of 210 drivers) will complete to month 18
- Drivers that attrition out will be placed into Group 1 (baseline) or Group 3 (shorter term)

Anticipated Outcome



Question: Can driver performance and safety improve over time with OBMS feedback and coaching?

Performance variables

- Performance data (e.g., speed, acceleration, lane deviation)
- Number of safety-critical events per miles traveled (e.g., lane departure events, distraction events, fatigued events)

Accounting for...

- Demographic data (e.g., gender, age)
- Environmental condition (lighting, traffic, roadway, weather, time-of-day)
- Seasonal effects

Random coefficient modeling

- Allows the analysis of multilevel data using a single regression model
- Frequently used in longitudinal study with data clustered across different levels
 - e.g., individual level, between-driver, and group levels
- Can account for non-normally distributed outcomes (e.g, safety critical events)
- Estimate the relationship of individual level predictors with the dependent variable while taking into account clustering

Random coefficient modeling

- **Level 1. individual growth model**, explains the performance change over time (baseline, feedback and withdrawal phase):

$$g(y_{ijt}) = \pi_{0ij} + \pi_{1ij} \times C_{ijt} + \pi_{2ij} \times X_{ijt} + \pi_{3ij} \times \mathbf{W}_{ijt} + e_{ijt}$$

- $g()$ represents the link function (e.g., Poisson distribution)
- C is the time variable of study participation (in months)
- X is the study phase (baseline, feedback, withdrawn)
- \mathbf{W} represents the matrix of environmental and seasonal factors

Random coefficient modeling

- **Level 2. driver-level model**, explains the between-subject difference within study groups (e.g., carriers, age, crash history, Z_{ij}):

$$\begin{aligned}\pi_{0ij} &= \beta_{00j} + \beta_{01j} \times Z_{ij} + r_{0ij} \\ \pi_{1ij} &= \beta_{10j} + \beta_{11j} \times Z_{ij} + r_{1ij} \\ \pi_{2ij} &= \beta_{20j} + \beta_{21j} \times Z_{ij} + r_{2ij} \\ \pi_{3ij} &= \beta_{30j} + \beta_{31j} \times Z_{ij} + r_{2ij}\end{aligned}$$

Random coefficient modeling

- **Level 3. group-level model**, explains the experimental group differences (e.g., baseline, long-term adaptation, short-term adaptation, V_j):

$$\beta_{00j} = \gamma_{000} + \gamma_{001} \times V_j + \mu_{00j}$$

$$\beta_{01j} = \gamma_{010} + \gamma_{011} \times V_j + \mu_{01j}$$

.....

Question: If driving performance improves, does it remain improved over time

Autoregressive integrative moving average (ARIMA)

- Time series model
- Accounts for seasonality and influence of change from interventions/withdrawals
- To observe if good driver performance persists even after feedback is removed

*Covariate:
Baseline group*

$$y_t - y_{t-k} = \varphi(y_{t-1} - y_{t-1-k}) + e_t - \theta e_{t-1} - \Theta e_{t-1-k} + \text{OBMS} + \text{CG}$$

*Safety critical
events/miles
traveled/month*

Seasonal differences

*When feedback was
turned on/off*



Question: How do the driver's opinions and attitudes towards the OBMS system and program change over time?

Questionnaires

- Baseline (before feedback is received)
 - Assessing driver's expectations of the OBMS system
- Feedback
 - Assessing driver's experiences with the OBMS system after they have received feedbacks.
- Withdrawal
 - system when feedbacks are removed.

Questionnaires

Data analysis

- Regression model to predict change (Δ) in response from one phase to the next
- Cluster analysis
 - to observe homogeneous groups of drivers based on their questionnaire responses

Summary

- Study is currently in the pilot phase
- Other areas being evaluated
 - Distinguishing between safe and unsafe drivers
 - Hours of Service: Electronic On Board Monitoring
 - Economics: Cost/benefit analysis of system implementation for carriers
- Other Issues
 - Epoch data only of safety critical events
 - No random events that can provide insights on how drivers may adapt to the system
 - Sampling biases: will need to account for statistically

Acknowledgments

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