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DRIVER COMPETENCE PERFORMANCE INDICATORS USING OTMR

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Abstract
The current practice for assessing driver competence performance is in-cab riding by driver managers. However, this paper investigates whether real-world driving data extracted from on-train monitoring recorders data (OTMR) can be used to assess the driver performance. A number of indicators were used to evaluate the drivers’ performance. These include: their use of the emergency bypass switch, the driver's reminder appliance as well as the driver’s reaction time.

A study case illustrated the applicability of OTMR data to estimate the proposed indicators, which suggests that the indicators can be useful in the driver management system in addition to the current indicators. Furthermore, the proposed indicators could be used to tailor the driver training schemes up to their individual needs and evaluate their effectiveness. They could even be used for improving driver competence performance and reducing crash involvement by revealing potentially detrimental driving performance.

1 Introduction

OTMR are used to collect data about the way trains are driven and the state of various train systems during its journey. Examples of data collected include power and brake controller position, driver acknowledgement of signalling system warnings, whether the doors are open and the operation of driver's reminder appliance and the emergency bypass switch systems. The OTMR data is currently used in:

- incident/accident investigation, for example, beside other sources, OTMR is reviewed for a double signal passed at danger (SPAD) incident at Esher on 25 November 2005 (DfT, 2007),
- automated train condition monitoring, for example, TAPAS condition monitoring system processes data recorded by OTMR to identify the required maintenance for trains (Green et al., 2001),
- automated driver assessment, for example, TAPAS and Churros process OTMR data to estimate a number of speed indicators such as the speed at which power notch 4 is selected when accelerating.

This paper extends the use of OTMR data in driver assessment to address another issue such as drivers’ use of safety systems as required by the Rule Book (GE/RT8000/TW5, 2014) as well as to estimate the driver’s reaction time.

2 Literature review

2.1 Human factor

Elms (2011) and Feldmann et al. (2008) suggested that human factors represent the central part in rail safety. However, RSSB (2009) identified that accidents are multifactorial; 2283 incident factors were the cause of the 292 events analysed (an average of 8 factors per event), reflecting the multi-causal nature of incidents.

In general, individual attitudes are reflected by compliance to procedures, effective thinking when facing an unforeseen situation, and management aspects for a good safety attitude (Sorensen, 2002). According to HSE (2009), human failures could be classified
into two categories, viz. human errors and violation. Both types may lead to an undesirable outcome. However, the first is when an intended decision or action deviates from an accepted standard and the latter is a deliberate deviation from rules, procedures, instructions and regulations. According to Flin et al. (2008), human errors are inevitable but all potential consequences need to be managed and mitigated by some non-technical skills (NTS) such as rule compliance, situation awareness and effective communication. The Office of Rail and Road (ORR, 2013) reported that one of the train operating companies linked the causes of its SPAD cases during 2010 to deficiencies in areas related with NTS such as lack of situational awareness and conscientiousness. Furthermore, following the introduction of non-technical skills training, Canadian Pacific Rail achieved a 46% reduction in human-caused incidents and recorded the lowest incident rate for class 1 railways in North America (ORR, 2012). Furthermore, a continuous assessment of drivers could have a positive impact on rail safety as individuals’ behaviour could be positivity influenced by being aware they are monitored (Wouters and Bos, 2000).

2.2 Train driver performance indicators
Despite the availability of massive source of data, i.e. OTMR, to date, there has been very little research to investigate train driver performance based on such data. Green et al. (2001) presented a number of performance indicators to facilitate the automatic analysis of driver performance. These indicators are: the speed at which power notch 4 is selected when accelerating, the percentage of time in a braking sequence that the driver selects brake step 3, the speed over Train Protection and Warning System (TPWS) grids approaching a Permanent Speed Restriction (PSR), the speed through a PSR as a percentage of the maximum speed and the mean speed when the automatic warning system (AWS) horn is received. The indicators are compared with average performance of the whole population of train drivers to assess an individual’s driving performance in relation to the cohort of drivers using TAPAS and Churros software. Furthermore, Churros software shows individual events when a driver has made an error such as wrong-side door release and TPWS brake demand. The indicators used by TAPAS and Churros reflect the important issues in driver assessment process. However, in this paper, another group of indicators are suggested to enhance the assessment of the train driver performance, for example, drivers’ use of safety systems as well as drivers’ reaction time. The following subsections introduce the proposed indicators and highlight the importance of each one.

2.3 Use of Safety systems
The Rule Book (GE/RT8000/TW5, 2014) emphasises the responsibility of the driver to set a number of safety systems such as Emergency Bypass Switch (EBS), Driver's Reminder Appliance (DRA) and TPWS soon after entering the driving cab, stopped at station, and leaving the driving cab stages. Even so, in some cases drivers may switch off a safety system, for example, the initial investigation of the incident near Doncaster on 2 October 2015 found that staff on-board locomotive 45231 turned-off its TPWS (Railway Herald, 2015).

Despite the importance of use of safety systems, there is no mechanism to continuously monitor the use of these systems during train journeys. A number of Train operating companies use in cab assessment to monitor drivers' operational usage of DRA (McCorquodale et al., 2002). In research level (RSSB, 2004), digital cameras were implemented to record driver’s action. Both previous techniques (in-cab assessment and in cab-observations) have their own merits in assessing the driver performance as they supplied comprehensive details about the driver performance. However, drivers may behave differently under observation, limiting the potential for independent driver
assessment. Add to that, the time and money cost for both techniques hinders their use for continuous monitoring.

The use of OTMR data alternatively offers a continuous cheap source for monitoring the use of safety systems without distracting drivers. Up till now, the data related to the safety systems in OTMR are only reviewed in case of incident or accident. Including the use of safety systems in the assessment of the train drivers may enhance the rail safety.

2.4 Driver reaction time

Driver’s reaction time has a potential effect on their competence level due to its practical implications. For example, the driver has to acknowledge safety system massages, such as TPWS horn within around 2.5 second to avoid an error leading to a braking application. Longer reaction time is also related to the driver fatigue level (Dorrian et al., 2007; Ji et al. 2004).

Furthermore, instance acknowledgment of safety system massages (i.e. driver acknowledges TPWS horn in less than 0.1 second) is also a bad practice as it could indicate unconscious response (Crick et al., 2004). Instance response also refers to “predict and act” behaviour which could have a further implication as discussed in Walker (2015).

Previous studies (e.g. Crick et al. 2004; Mcleod et al., 2004) implemented a number of methods such as in-cab observation, retrospective analysis and focus groups to estimate the average reaction time of drivers. Despite the effectiveness of these methods in estimating the average value of reaction time and highlight the common issues, they do not show the variation in the individual driver reaction time. Furthermore, they are not designed to introduce a continuous indicator for driver reaction time. To avoid this shortage, this research has implemented OTMR data to estimate the driver reaction time. By doing so, the extreme values, i.e. instance and late responses for each driver will be identified, hence required driver training could be conducted.

3 Method

Figure 1 presents the three steps used to estimate the proposed indicators. In the OTMR raw data stage, an algorithm has been developed to convert OTMR data format to CSV files format. The main purpose is to be able to handle multi files in R software and save tremendous time of data analysis and plotting. The initial handling of data stage examines data types and format. Furthermore, closer inspection of data shows a number of deficiencies that can affect the reliability of data analysis such as missing values across all the variables and unknown or unexpected character encoding. To automate the correction of data type and format, an R script is written and checked against a manual calculation using Excel. Figure 2 gives an example of changing “Time” and “Relative Time” format to facilitate the use of both variables in the data analysis stage.
In this paper, EBS and DRA are chosen as an example of safety systems that the driver has to apply. The use of EBS is monitored for the entire journey, whereas the driver has to set DRA at every red signal experienced, as required by the Rule book (GE/RT8000/TW5, 2014). Furthermore, the driver reaction time is evaluated based on the time taken to cancel the TPWS horn as a proxy measure, i.e. horn and acknowledgment.

In OTMR data analysis stage, a number of OTMR channels are used to extract the relevant scenarios as presented in Table 1. For example, zero speed periods are calculated based on the train speed. The door release channels are, then, examined to determine if the train is stopped in a station or in front of a red signal, followed by a check of DRA channels.

<table>
<thead>
<tr>
<th>Situation</th>
<th>OTMR Channels</th>
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</thead>
<tbody>
<tr>
<td>Use of EBS</td>
<td>• EBS channels,</td>
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<tr>
<td></td>
<td>• Relative time channel.</td>
</tr>
<tr>
<td>Consistent use of DRA when facing a red signal</td>
<td>• DRA channels,</td>
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<td></td>
<td>• Speed channel,</td>
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<tr>
<td></td>
<td>• Door-release channels.</td>
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<tr>
<td></td>
<td>• Relative time channel.</td>
</tr>
<tr>
<td>Reaction time</td>
<td>• TPWS channels,</td>
</tr>
<tr>
<td></td>
<td>• Relative time channel.</td>
</tr>
</tbody>
</table>

Table 1 Detecting different situations using OTMR channels.

An algorithm is developed in R environment to extract the required scenarios from OTMR data files and to estimate the use of DRA and EBS in addition to calculating the driver reaction time.

4 Study case

4.1 Data
The OTMR data files used in this paper were supplied by Southern Railway. They are for the same route and the same day but different drivers to eliminate the impact of route conditions. The steps explained in Section 3 are implemented to estimate the proposed indicators.
4.2 Results and discussion

4.2.1 Use of EBS and DRA

The analysis of OTMR files shows that 100% of OTMR data files comply with EBS use. Figure 3 shows the number of the red signals drivers experienced during each journey and the corresponding use of DRA, indicating that all drivers complied with the rules except for Journey 6. The Journey 6 driver experienced 4 red signals but set the DRA 3 times. Consequently, a further analysis has been carried out to confirm these findings. Figure 4 presents the train speed, door release (left hand side and right hand side of cab in use), and DRA use of Journey 6. The driver did not set the DRA for the first red signal he/she experienced, indicated with the red circle in Figure 4, whereas set it for the rest of red signals, shown in the blue circle in Figure 4. The driver also used the DRA when stopped at station 10, suggesting that the driver faced a red signal at the end of the platform. It should be noted that, the data from OTMR journey files can confirm the driver set the DRA in case of red signal at the end of the platform as required by the Rule Book. However, if the driver does not set the DRA in case of a red signal at the end of the platform, OTMR channels cannot detect this error as such scenario cannot be identified by OTMR data only.

Figure 3 Number of red signals drivers experienced during each journey and the corresponding use of DRA.
4.2.2 Driver’s reaction time

Box plots, presented in Figure 5, are used to show driver’s reaction time variability during a journey in addition to the variation among drivers. All the drivers are in compliance with the max reaction time (i.e. less than 2.2 seconds) as the maximum value is around 1.4 seconds as presented in Figure 5. In general, Figure 5 shows a wide variation in reaction time patterns, but it is observed the majority of the drivers have the reaction time below 0.6 second, that is lower than 0.6-0.9 seconds average reaction time suggested by other studies (Scott, 2008; Crick et al.2004; Mcleod et al., 2004). Furthermore, 8 drivers out of 12 have a minimum of zero reaction time (less than 0.1 second, the interval of OTMR data recording) at least once during their journey. These values suggested that the drivers may consider acknowledgment of TPWS horn as a high priority task to avoid TPWS brake demand which is classified as an error. However, very quick response could indicate the unconscious cancelling of warning signs as stated by RSSB (2004) and, in such case, there is a need to be addressed by a suitable training.

In terms of individual driver, for example, in journey 10, the box plot is extremely short, indicating that the driver’s reaction time is nearly constant with a value around 0.2 seconds. In contrast, Journey 2 has a longer box plot, suggesting a wider variation in the driver reaction times during the journey. Furthermore, the minimum reaction time is equal to zero whereas the highest reaction time is equal to 1.4 seconds.
5 Conclusions
This paper proposed a number of driver performance indicators based on real-world driving data extracted from OTMR data files. The use of safety systems such as EBS and DRA and the driver reaction time are proposed.
The study case illustrated the applicability of the proposed indicators. Assessing drivers performance based on real-world driving data offers a fair method for continuous assessment of drivers performance under real life conditions. Besides, it could be used to develop the required training scheme for drivers based on their driving data.

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