

Tenth International Conference on Managing Fatigue Abstract

Circadian rhythmicity and Risk Index validation in railway traffic control

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Problem

1 This paper describes the findings of a data-driven fatigue research project at Belgian
2 railway traffic control. First, we estimate the circadian rhythmicity in registered human
3 errors. Second, we statistically evaluate the validity of the Risk Index tool (Spencer et
4 al., 2006) through a multivariate Tobit regression analysis of its impact on these human
5 errors. The regression model also considers factors not taken into account by the Risk
6 Index methodology. As such, we extend and complement previous research on the validity
7 of the Risk Index by Greubel et al. (2010) and Greubel and Nachreiner (2013).

Method

8 In order to estimate the circadian rhythm in hourly error occurrence, and following
9 Folkard et al. (2006), we perform several cosinor regressions on human errors in rail-
10 way traffic control. Error occurrence is corrected for exposure by taking into account
11 traffic volumes.

12 Next, a multivariate Tobit model empirically examines the relationship between the
13 exposure-corrected human errors and the Spencer et al. (2006) Risk Index scores. By
14 applying a Tobit regression, we account for zero values in error occurrence. The re-
15 gression model additionally considers variables capturing industry-specific settings, age,
16 gender, part-time work and day-of-week (i.e., factors which are not taken into account
17 by the Risk Index methodology).

18 In close cooperation with railway experts from Infrabel, the state-owned company running
19 the Belgian railway infrastructure, we analyze real-life data from 11 computerized Traffic
20 Control Centers. We retrospectively analyze a unique full year dataset, containing more
21 than 11,000 work shifts. The signalling and dispatching work in the Traffic Control
22 Centers is standardized through non-overlapping 8-hour work shifts, starting at 06:00,
23 14:00 and 22:00 (i.e., early, late, and night shift). However, local management has the

24 authority to organize and adapt work shift patterns in functions of their needs (e.g.
25 team composition, direction and speed of shift rotation, distribution of rest days). Work
26 schedule risk is assessed for each individual traffic controller with actual staff schedules.
27 We complement the obtained Risk Index scores with intra-company data (e.g. traffic
28 volumes, traffic controller age and gender). The dataset is further enriched with the
29 human errors detected by the traffic control system (relatively frequent but non-critical
30 task errors, such as erroneous ordering of signal commands). The human errors are not
31 identified at individual level, but at team level (i.e., the group of traffic controllers present
32 in the control center during the work shift). For the purpose of the Tobit regression, data
33 is further aggregated by each 8-hour work shift. Data is collected, verified, and validated
34 by a custom-developed Business Intelligence tool.

Results

35 The cosinor parameter estimates are all statistically significant, with an acrophase es-
36 timate (circadian peak) around 2 am. This estimate proves robust to alternative data
37 aggregations (e.g. with errors calculated per shift instead of per hour, see Folkard et al.,
38 2006).

39 The multivariate Tobit regression reveals a positive and highly significant effect of the
40 average team Risk Index on human error occurrence. Control variables reflecting opera-
41 tional conditions (such as the percentage of automatically commanded signals) are also
42 significant. There is no significant impact of average team age, gender (percentage of
43 male traffic controllers) or part-time work (percentage in the team). Finally, day-of-week
44 dummy variables exhibit varying parameter signs and significance levels. Tobit marginal
45 effects indicate that, all other things being equal, the probability of making at least one
46 error is highest on Saturdays (+ 6% compared to Mondays), and lowest on Tuesdays,
47 Wednesdays and Thursdays. Regression results are robust to changes in model specifica-
48 tion.

Discussion

49 The estimated 24-hour rhythm, exhibiting an acrophase at 02:00 hours, is closely aligned
50 with previous research examining circadian employee performance (Folkard and Tucker,
51 2003, circadian low at 03:00) or the risk for accidents and injuries (Folkard et al., 2006,
52 acrophase around midnight).

53 Our Tobit regression result validates the Risk Index in a real-life setting, and therefore
54 extends previous web survey - based research on the validity of the Risk Index by Greubel
55 et al. (2010) and Greubel and Nachreiner (2013). Moreover, following a suggestion by
56 Greubel and Nachreiner (2013), we analyzed a potential ‘day-of-week’ effect and found a
57 significant impact on the exposure-corrected error levels.

58 Also, in line with the recommendation by Dawson et al. (in press) to apply biomathemat-
59 ical models in a ‘post-implementation surveillance mode’, our custom-developed Business

60 Intelligence tool allows for a further (non-statistical but operationally intuitive) probing
61 of the data by railway experts.

Summary

62 The present study is part of an ongoing fatigue risk research project, performed in close
63 collaboration with the Belgian railway infrastructure company Infrabel. Applying cosinor
64 rhythmometry and Tobit regressions, our analysis not only estimates circadian rhythmic-
65 ity in human error, but also validates the Risk Index Spencer et al. (2006) under real-world
66 circumstances. As suggested by our regression results, an enhancement of the Risk Index
67 to account for ‘day-of-week’ effects could further reinforce the accuracy of the tool. Fi-
68 nally, our research also aims to bridge the gap between theory and practice, by deploying
69 a Business Intelligence tool for ex-post Risk Index analysis.

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