

RELATIONSHIPS BETWEEN THE MOTORCYCLISTS' BEHAVIOURAL PERCEPTION AND THEIR ACTUAL BEHAVIOUR

Darja Topolšek¹, Dejan Dragan²

¹*Dept of Technical Logistics, University of Maribor, Celje, Slovenia*

²*Dept of Quantitative Modelling in Logistics, University of Maribor, Celje, Slovenia*

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Abstract. The paper deals with studying the relationships between the motorcyclists' thinking about proper behaviour and their actual behaviour in the traffic. The impact of some control variables, such as riders' age, experience, driving history, and engine cubature, on actual behaviour, is also addressed here. For the purpose of research, two additional questionnaires were applied besides the well-known Motorcycle Rider Behaviour Questionnaire (MRBQ). To examine the causal relations between all-important latent factors present in this study, the structural equation model was designed. Exploratory and confirmatory factor analyses were also engaged in the analysis and the statistical modelling process. The results show that the higher awareness about alcohol danger and benefits of protective equipment and helmet can noticeably contribute to the bigger traffic safety. Besides, from the results is evident that the control variables are in most cases also significantly interrelated with the actual behaviour factors. The findings of this research could be important for the planning of better traffic safety strategies for the motorcyclists to decrease the fatalities and related costs and traumas.

Keywords: motorcycle rider behaviour questionnaire (MRBQ); motorcyclists' behavioural perception; traffic safety; alcohol; factor analysis; structural equation modelling.

DRAFT

Introduction

If compared with the huge amount of research, conducted in the field of traffic safety about four-wheeled vehicles, relatively little research was carried out in the area of traffic safety of Powered Two Wheelers (PTW). This is particularly true for the motorcycles-related studies. Such a gap could wrongly imply that the motorcycle accidents are negligible, but it is well-known that they are not. On the contrary, the number of traffic accidents with the motorcyclists involved is significant and is even on the rise, while, on the other hand, the total number of traffic accidents is decreasing (OECD 2014).

Riders of powered two-wheelers in the European Union are one of the most vulnerable groups of road users (Šraml *et al.* 2012). They are quite often involved in road accident and, unfortunately, in many cases can be seriously injured or even killed. Indeed, some studies of motorcycle accidents reported that approximately 96% of motorcyclists involved in traffic accidents are at great risk to suffer certain injury (Hurt *et al.* 1981). Moreover, other studies show that even in 50% of such accidents,

serious injuries or even death of the rider occurred (Diamantopoulou *et al.* 1995).

To offer better traffic safety strategies for the motorcyclists, it is very important to understand the main factors leading to accidents followed by tragic consequences (heavy injuries or death). Unambiguously, riders' errors, decision failures, and the perception failures are some of the most noticeable issues, besides alcohol consumption, driver's age, and wearing protective clothing and a helmet – the Motorcycle Accidents In Depth Study project (MAIDS 2009).

The MAIDS project revealed that 35.9% of all motorcyclists caused an accident due to the rider's error, 13.2% of them made decision failures, while 8% of the accidents were a result of perception failures. Almost 15% of motorcycle-related accidents ended with a fatal outcome, despite the wearing of the helmet in 98.6% of all cases, and regardless of relatively insignificant presence of alcohol or drugs (only in 3.3% cases). The corresponding study also discovered that the number of accidents was related to the age of the rider, the most representative age group being from 26 to 40 years.



In general, many studies discuss the interrelations between alcohol consumption, protective clothing, helmet wearing, and rider's behaviour on one side, and more or less tragic consequences of road accidents on the other. Alcohol is proved to be related to motorcycle accidents (i.e. Turner, Georggi 2001; Villaveces *et al.* 2003; Williams 2006). As opposed to other road users, riders are more sensitive to the effects of alcohol since they also have to worry about balance and the coordination of the motorcycle. Lin and Kraus (2009) proposed that a zero limit of the blood alcohol concentration should be obligatory since only this legislative measure could ensure the adequate coordination and balance when operating two-wheeled vehicles.

European Union has a standard, which requires the mandatory use of motorcycle gloves, boots, jackets and pants (EC 1989). Several studies reveal that protective clothes reduce the risk of soft tissue injuries, but do not protect against serious injuries (Otte *et al.* 2002). Some researchers also confirmed that the use of protective clothing is associated with the purpose of riding (De Rome 2006; Watson *et al.* 2008). In addition to that, it was discovered that the club members' riders more frequently use motorcycle boots and pants than the non-members riders (De Rome, Stanford 2006). De Rome *et al.* (2011) also concluded that riders without protective clothing do not really believe in the practical value of such protection against injury.

The importance of the helmet and its role in preventing head injuries and deaths has been discussed by many scholars (Evans, Frick 1988; Wilson 1989; Gabella *et al.* 1995; Rowland *et al.* 1996; Lin *et al.* 2001; Liu *et al.* 2004). Many studies have identified head injuries can contribute to an astonishingly high percentage of motorcycle accidents with a fatal outcome, even up to 60% (Azhar *et al.* 2014).

The usage of helmets is mandatory in the European Union countries, which results in the high share of their actual use. Therefore, generally, riders wear a helmet due to legislative measures, but their personal opinion about the helmet use is questionable. As it turns out, only a few studies examined the use of helmets in conjunction with the riders' personal opinion about the helmets' usefulness (Rutter *et al.* 1998; Pileggi *et al.* 2006; Oginni *et al.* 2007; Germini *et al.* 2009; Ranney *et al.* 2010). The majority of these studies concluded that riders with no helmets do not truly believe in their protective value. Ranney *et al.* (2010) concluded that not-always-helmeted and always-helmeted motorcyclists more or less believe in the priceless value of wearing the protective gear and helmet. On the other side, it was also discovered that non-helmet motorcyclists also more frequently show preferences to a riskier behaviour (Chen 2006; Brown *et al.* 2011; Haworth *et al.* 2009).

Statistics for Slovenia show, similarly as in other European Union countries that the motorcyclists are at greater risk to be severely injured or even killed in traffic accidents if compared to the car drivers. In Slovenian local terminology, all types of unusual events, which occur on Slovenian roads, are treated, perceived and officially

recorded as 'traffic accidents', regardless their nature or possible consequences. For instance, the events with motorcyclists involved, such as falls, slips, collisions, crashes, and other events, are all addressed as traffic accidents, irrespective of their intensity and potential damage, possible injuries, or even tragic consequences.

For Slovenian traffic safety is also discovered that the deceased riders represent a large proportion of the average number of all traffic losses, which appeared to be 63 fatalities per million inhabitants during the last decade (Zanne *et al.* 2013). According to the official statistics, one of the reasons for such tragic consequences is also interrelated with the very bad condition of most Slovenian roads, which are in many cases deteriorated and insufficiently maintained. In addition to that, the driving culture of road users is still inappropriate, while the traffic volume rapidly increases during the last years (Policija 2014).

Another important fact for Slovenia is that about 600 traffic accidents happened with motorcyclists engaged during the year 2013 (Policija 2014). On one side, this means only 2.5% of all traffic accidents in Slovenia, but, on the other hand, almost 20% of the motorcyclists were heavily injured or deceased in these accidents (Policija 2014). Furthermore, the accidents in 43% of all cases happened because of the riders, among whose 7% were influenced by the alcohol, while over 95% of them carried a (mandatory) helmet (Policija 2014). As in the entire European Union area, the most representative age of the riders involved in the accident belonged to the group from 26 to 40 years.

A deeper examination of traffic accidents in Slovenia shows a quite worrying picture, if compared to the majority of European Union countries. According to the 2nd Road Safety PIN Report (Jost *et al.* 2008), published by the European Transport Safety Council (ETSC) in 2008, Slovenia 'won' the shameful first place regarding the PTW motorcyclists' deaths per billion travelled kilometres in the year 2006 (Šraml *et al.* 2012). It had the highest number of such kinds of deaths (more than 350, while the average in European Union was 'only' 86). Another very shocking fact for Slovenia is that the risk of a rider being killed in a traffic accident is on average more than 50 times higher than the corresponding risk for a car driver (the European Union average 'only' 18 times) (Jost *et al.* 2008; Šraml *et al.* 2012). In addition, as reported by United Nations Economic Commission for Europe – Transport division (UNECE 2015), in the past decade Slovenia had a significantly higher ratio of injured riders per number of registered motorcycles than the majority of other European Union countries. For instance, this ratio was 647/41600 (about 1.5%) in year 2008, while the European Union average was below 1% (UNECE 2015, Eurostat 2015). All these facts indicate that the riders are obviously more jeopardised on Slovenian roads and can be more likely involved or even killed in an accident than their counterparts in most other European Union countries.

In order to contribute to the maximal possible reduction of the motorcyclist accidents, it is certainly use-

ful to get a certain insight into the riders' thinking and perception about his/her driving behaviour and the attitude to the alcohol, usefulness of the protective clothing, and assets of the helmet. In this spirit, the main goal of the paper is to investigate the relationships between the rider's perception about the alcohol, clothing, and helmets on one side, and his/her thinking about the safety and behaviour on the other. In other words, do the drivers behave in the same way as they think they actually do? The latter means the issue, if believing about the danger of drinking alcohol and usefulness of wearing protective clothing and helmet truly result in the safer behaviour of the riders. On the other side, this might be only self-deception, and the truth is completely different from the riders' personal perception. Since practically none similar researches were detected in the existing literature, we believe that the filling of this gap might be one of the major contributions of this paper.

For the purpose of research, an interview was applied among the randomly chosen group of Slovenian motorcyclists. The basis for interviewing was well-known Motorcycle Rider Behaviour Questionnaire (MRBQ), firstly introduced in the study (Elliott *et al.* 2007), and later additionally tested in the other studies (Sakashita *et al.* 2014; etc.). Its main purpose is to measure the riders' behavioural characteristics, such as control and traffic errors, use of safety equipment, and speed violations, as well as to identify how are the latter related to the crash risk (Elliott *et al.* 2007). The MRBQ consists of 43 items related to the safe or dangerous behaviour of the riders and is usually measured on six or seven points Likert scale. The latest MRBQ studies have discovered that the resulting factor model can be expressed by four significant factors, which are speed violations, errors, safety equipment, and stunts (Sakashita *et al.* 2014).

In our case, we used only the 5-point Likert scale for the MRBQ items due to the local characteristics. The details about the structure of these items can be found in the scholarly literature (Elliott *et al.* 2007; Özkan *et al.* 2006; Sakashita *et al.* 2014).

Besides the MRBQ, two additional questionnaires were given to the riders for the fulfilment. The first one (questionnaire Q1) was interrelated with the riders' subjective thinking (perception) and was composed of the alcohol based, helmet based, and clothing based items (measured indicator variables). The second one (questionnaire Q2) contained the control single-items, which are age, years of driving, motorcycle engine volume (cubature), mileage per year, safety driving trainings, and the number of experienced accidents.

After the completed interview, the descriptive statistic of the collected data was investigated at first. Then, the preliminary Exploratory Factor Analysis (EFA), Confirmatory Factor Analysis (CFA) and the Structural Equation Modelling (SEM) procedure were applied. These techniques are very powerful statistical modelling tools from the field of multivariate statistical analysis (Kline 2015; Hoyle 2014; Hair *et al.* 2009; Byrne 2009). They are used in many scientific areas, such as in the social sciences, psychometrics, management, econom-

ics, and operations research, as well as in the natural sciences, engineering, marketing research, educational research, and tourism. Occasionally, they have been also used in the traffic safety research, for example, in the studies (Mattsson 2012; Lajunen *et al.* 2004; Elliott *et al.* 2007; Özkan *et al.* 2006).

When the (unmeasured) latent factors and their loadings were adequately extracted from the given items' data by the means of EFA, the preliminary factor model was a result. The latter represented a useful guideline for the further analysis, by which the so-called structural equation model (SEM model) was afterward calculated in two stages.

In the first stage, the so-called measurement part of the model was derived by the means of CFA. Subsequently, in the second stage, the so-called structural part of the model was also constructed, which enabled us to finish the design of the overall SEM model. The latter provided the means to analyse the causal relations between all factors involved in the analysis (MRBQ multiple-item based factors, Q1 multiple-item based factors, and single-item factors of the questionnaire Q2). All the calculations were conducted in the program package IBM® SPSS® 21 and its extension Amos™ (Byrne 2009; Arbuckle 2012).

1. Conceptual Framework, Survey and Hypotheses

1.1. The Conceptual Framework

Fig. 1 shows the conceptual framework with the hypothesized model. Forty three (43) items of the MRBQ questionnaire are symbolized by variables M_i , $i = 1, \dots, 43$, while the 11 items related to Q1 are denoted by: A_i , $i = 1, \dots, 4$ for the alcohol; C_i , $i = 1, \dots, 4$ for the motorcycle clothing (protective jacket, trousers, gloves, and boots) and H_i , $i = 1, \dots, 3$ for the helmet. The meaning of the latter will be explained in the sequel. Control items of the questionnaire Q2 are also shown in Fig. 1, and so are four MRBQ factors (speed violations, errors, safety equipment, stunts). It is supposed that the further statistical analysis will clearly give us the three factors related to the Q1, which can be entitled by: alcohol, helmet, and clothing (Q1 factors in Fig. 1). We assume that these Q1 factors and the control single-item Q2 factors definitely have a certain impact on the MRBQ factors. Their influences could be marked with the hypothesized paths containing $9 \cdot 4 = 36$ hypotheses $\hat{H}_i$, $i = 1, \dots, 36$, each representing the assumed singular impact of Q1 and Q2 factors on the MRBQ factors. Since the number of 36 hypotheses is quite big, we decided to simplify them to the lower number of only seven grouped hypotheses. This way, the following comprised hypotheses have been applied (Fig. 1):

$\hat{H}_{ALC} = \{\tilde{H}_i\}$, $i = 1, \dots, 4$ – impact of alcohol factor on MRBQ factors;

$\hat{H}_{CLO} = \{\tilde{H}_i\}$, $i = 5, \dots, 8$ – impact of clothing factor on MRBQ factors;

$\hat{H}_{HEL} = \{\tilde{H}_i\}$, $i = 9, \dots, 12$ – impact of helmet factor on MRBQ factors;

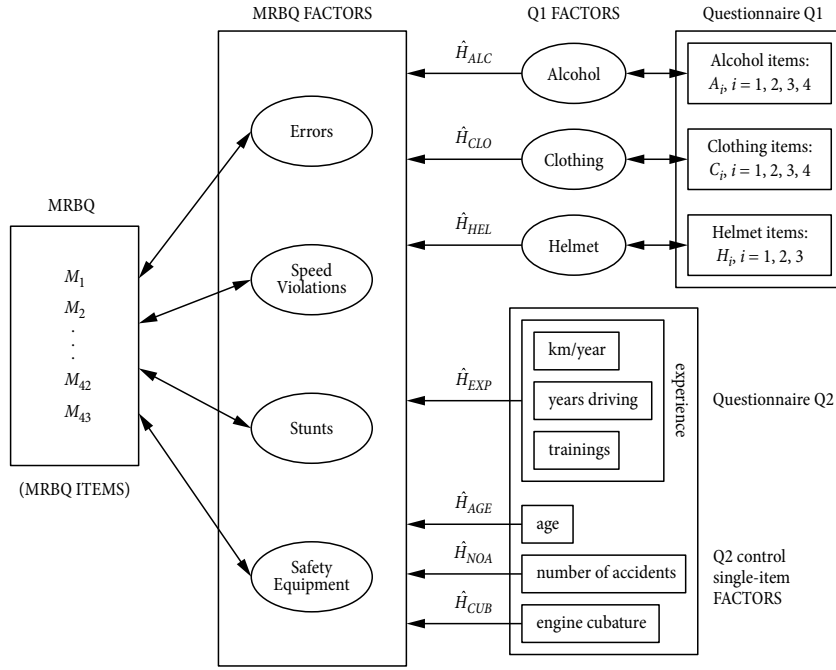


Fig. 1. The conceptual framework (Q1 factors – motorcyclists' subjective opinion about proper behaving; MRBQ factors – motorcyclists' actual behaving)

$\hat{H}_{AGE} = \{\tilde{H}_i\}, i = 13, \dots, 16$ – impact of age factor on MRBQ factors;

$\hat{H}_{EXP} = \{\tilde{H}_i\}, i = 17, \dots, 28$ – impact of experience on MRBQ factors (comprises years of driving [km/year], safe driving trainings);

$\hat{H}_{NOA} = \{\tilde{H}_i\}, i = 29, \dots, 32$ – impact of accidents' factor on MRBQ factors;

$\hat{H}_{CUB} = \{\tilde{H}_i\}, i = 33, \dots, 36$ – impact of engine cubature on MRBQ factors. (1)

From Fig. 1 can be also depicted that the driving experience of the riders' was combined with three characteristics, namely the years of driving, mileage per year [in kilometres], and the frequency of safe driving trainings.

1.2. Further Explanation of the Settled Hypotheses

In general, motorcycle riders agree that a definition of safe rider includes the awareness of the danger of high-speed driving and the perception of the importance of the maximal focus on the driving within some reasonable limits. However, the riders often have difficulties with defining or expressing these terms (awareness, perception, focus, limits, and so) (Watson *et al.* 2007). Some riders also agree that a safe rider should always wear good protective clothing (Watson *et al.* 2007). However, for instance, do they actually wear the motorcycle clothing as they think they should do, or they are just fooling themselves and the others?

Nevertheless, the thinking about a certain thing is not always the same as truly doing this thing. To discover the possible discrepancy between riders' thinking and their truly performing/behaving, the hypotheses in the previous section could be also defined in the following way:

$\hat{H}_{ALC} = \{\text{motorcyclist, who is aware of the danger of alcoholic drinking, is a safer rider}\};$

$\hat{H}_{CLO} = \{\text{motorcyclist, who is aware of the benefits of protective clothing usage, is a safer rider}\};$

$\hat{H}_{HEL} = \{\text{motorcyclist, who is aware of the benefits of helmet usage, is a safer rider}\}$

$\hat{H}_{AGE} = \{\text{older motorcyclists are safer riders}\};$

$\hat{H}_{EXP} = \{\text{more experienced motorcyclists are safer riders}\};$

$\hat{H}_{NOA} = \{\text{motorcyclists, who have had a traffic accident, are the safer riders}\};$

$\hat{H}_{CUB} = \{\text{motorcyclists, who drive more powerful motorcycles, are more dangerous riders}\}. (2)$

When defining the precise interpretation of these hypotheses, the help offered by the interviewed riders was very useful. Therefore, based on their answers and comments, it was supposed that the higher level of awareness of the importance of helmets, clothing, and non-drinking alcohol encourages the riders to drive more carefully. Similarly, their personal opinion about the age, experience, engine cubature and frequency of already occurred accidents was considered not only in the hypotheses' construction, but also in the questionnaires' design.

1.3. Variables and Design of the Questionnaire Q1

The meaning of item variables of the questionnaire Q1 is shown in Table 1. As mentioned, this questionnaire is designed to recognize the subjective opinion of the motorists about the danger of alcohol consumption, the benefits of using the motorcycle helmets, and the usefulness of the protective clothing, respectively. For measur-

Table 1. Subjective opinion based items of the questionnaire Q1

Item	Measure
Alcohol items	
A ₁	not consuming alcohol while driving
A ₂	drinking the alcohol increases the likelihood of participation in a traffic accident
A ₃	motorcyclists are more sensitive to the effects of alcohol compared to other drivers since they also have to worry about the balance
A ₄	motorcyclists are more sensitive to the effects of alcohol compared to other drivers since they also have to worry about the coordination of the motor
Clothing items	
C ₁	motorcyclists are involved in a car accident since the car drivers did not see them, or they have seen them too late to avoid the collision
C ₂	protective clothing reduces the damage of the skin and soft tissues
C ₃	protective clothing reduces the deep and extensive injuries
C ₄	protective clothing reduces the time needed for the recovery after the injury happened
Helmet items	
H ₁	helmet saves lives and prevents the damage of the brain/skull
H ₂	helmet use reduces the severity of the neck injuries
H ₃	it is essential to have well-attached helmet on the head since it can fall off in the case of an accident

ing the Q1 variables, the Likert ordinal scale from 1 to 5 was applied (1 – completely disagree; 5 – completely agree). The questions related to the clothing were meant in the spirit of better visibility (fluorescent clothing) since the car drivers have more difficulties to see the riders in dark (un-fluorescent) clothing.

1.4. Design of the Questionnaire Q2

This questionnaire was designed to evaluate the effect of certain control variables on the behavioural factors, which are related to the motorcyclists' crash risk. After careful consideration, we have decided that the control measures age, years of driving, motorcycle engine cubature, mileage per year, safety driving trainings, and the number of experienced accidents, would be most appropriate for further analysis (Fig. 1). The sample characteristics of measured scale of Q2 control items are shown in Table 2.

1.5. Participants and the Execution of the Interview

All the needed data were collected over a 3-week period in the fall of 2014, and this collection was carried out by the means of online surveys. The latter were randomly sent to the most significant motorcycle clubs across Slovenia, which means that we can generally assume that the collected sample was random, unbiased and representative. After the completed survey, 205 fully com-

Table 2. The sample characteristics of measured scale of Q2 control items

Items/measures	Measured scale
Age	1 – below 20 years 2 – from 20 to 29 years 3 – from 30 to 39 years 4 – from 40 to 49 years 5 – from 50 to 59 years 6 – above 59 years
Years of motorcycle driving (total years holding the driving licence)	1 – less than 1 year 2 – from 1 to 2 years 3 – from 2 to 5 years 4 – from 6 to 10 years 5 – more than 10 years
Engine capacity (cubic centimetres)	1 – less than 500 cc 2 – from 501 to 1000 cc 3 – from 1001 to 1500 cc 4 – more than 1500 cc
Kilometres/year	continuous
Number of accidents	polytomous
Safe driving trainings	1 – each year 2 only once (when passing the driving test) 3 – never

pleted questionnaires were received, which have been afterwards included in the further research. The final sample encompassed 86.3% males and 13.7% females. The riders' age distribution is given in Table 3. 34.6% of all participants did not have motorcycled traffic accidents while 22% of them were involved in more than three accidents.

The percentage (64.4%) of observed riders already involved in any kind of accident (including lighter falls, slips, etc.) seems significantly larger than expected, but there are reasons for that. Firstly, as it was already mentioned in the introduction section, Slovenian roads are quite dangerous for the riders to be involved, injured or even killed in an accident. Namely, as it was estimated for the year 2008, about 1.5% of all motorcyclist population suffered the consequences of injuries of some type.

Secondly, there is a considerably high percentage of individual falls and slips, which are never reported. Specifically, as the Police, medical institutions, and motorcycle clubs claim, the riders often do not report 'minor' individual accidents (without injuries) in which no other vehicles are involved (falls, slips, and so). Most likely, it was the same for our group of riders, who probably treated such unreported minor events as an accident as well, when they filled in the questionnaires.

Thirdly, as found by the authors (Šraml *et al.* 2012), many PTW accidents occur on the regional state roads with low traffic volume, where PTW riders try to exploit wanton driving and provoking driving capabilities. Probably many of such accidents, particularly those with no serious consequences, are also not reported. Bearing all this in mind, we believe that the accident involvement characteristics of the sample are in accordance with the characteristics of the entire population, so the corresponding sample bias is within the reasonable level.

Furthermore, careful examination of the literature showed that there are practically no studies available which would address the frequency of all earlier PTW riders' accidents in other European Union countries. To the best of our knowledge, the Continental mobility study from the year 2013 is the only such study. In this study, the recorded accident involvement is similarly high for the German riders, as it was in above-mentioned case of Slovenian motorcyclists (Continental AG 2013). Explicitly, this means that more than half of all German motorists (59%) were already involved in some accident of any kind during their riding history (Continental AG 2013). On the other side, some studies also reported the percentage of the riders with at least one earlier accident involvement, but the related survey questions were focused on the accidents occurred only during the past year (for example, 11% of riders already involved in an accident during the previous 12 months were detected in the study Elliott *et al.* 2007).

Table 3. The riders' age distribution

Age range [years]	Percent of the sample
40–49	34.3
50–59	33.5
30–39	20.0
> 59	6.3
20–29	5.9

2. Methods, Analysis and Modelling

2.1. Used Methodology

Fig. 2 illustrates the methods, which were used in the analysis and modelling procedure. At first, the descriptive statistic of the measured data was investigated with emphasis on the analysis of normality. Afterwards, the

EFA was applied as a prior statistical technique to identify the nature of the latent factors (constructs) and to estimate their indicator items' loadings (Hair *et al.* 2009). This also enabled us to get a preliminary insight into the relationships between the observed, measured items and the corresponding unmeasured latent constructs.

The two-stage design of the structural equation model was the next step in our statistical analysis. In the first stage, the CFA was conducted, which exposed the EFA based theoretical factor model structure to the process of evaluation of measurement theory. This way, the quality of the presumed theoretical factor model fit to the real data was inspected (Hair *et al.* 2009). When the confirmatory measurement test of the factor model was successfully carried out, the measurement part of the SEM model appeared as a final result of CFA (Hair *et al.* 2009).

In the second stage, the structural part of the SEM model was also derived by the means of the SEM modelling procedure. The latter represents a very advanced statistical tool, which combines the factor analysis and the multiple regression analysis into the comprehensive modelling technique (Hoyle 2014; Dragan, Topolšek 2014). It also characterizes a generalization of causal path modelling and deals with the analysis of covariance structures (Hoyle 2014; Dragan, Topolšek 2014).

After the derivation of the overall SEM model, the Goodness Of Fit (GOF) measures was also examined, which enabled the testing of model validity and adequacy. The developed SEM model was afterward used to expose the causal directional relations between all factors involved in the study (MRBQ factors, Q1 factors, and Q2 single-item factors). All further details about the methodologies used in our study can be found in the scholarly literature (Kline 2015; Hoyle 2014; Hair *et al.* 2009; Byrne 2009; Mulaik 2009; Raykov, Marcoulides 2006; Timm 2002).

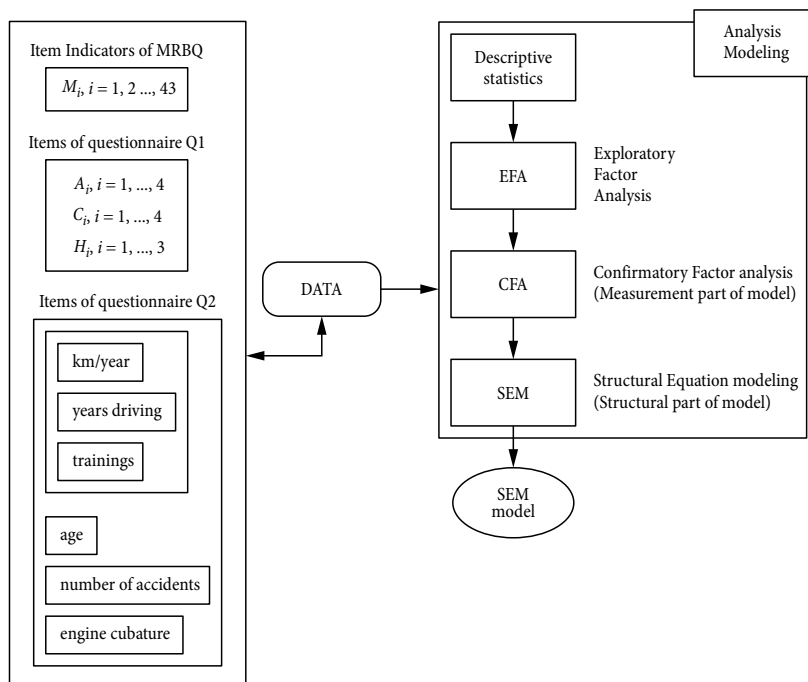


Fig. 2. The block diagram of the used methodology

2.2. Estimation Methods

For the purpose of parameter estimation in the CFA and SEM procedures, a variety of different estimation methods have been developed in the last decades (Kline 2015; Hoyle 2014; Hair *et al.* 2009; Byrne 2009). The most common is the Maximum Likelihood (ML) estimator, which is usually used when the data are at least approximately normally distributed. Besides, some other typical estimation methods are: un-weighted least square methods, generalized least squares methods, asymptotically distribution-free methods, Bayesian estimation methods, and so (Kline 2015; Hoyle 2014; Byrne 2009).

In our case, the ML method was used in the CFA and SEM since the major indicator variables (those of questionnaires MRBQ and Q1) have had only slight non-normal distribution. The excuse for applying ML estimator is based on the conclusions of several other studies. Namely, it is reported that the ML estimator gives the suitably accurate estimated parameters, if the ordinal indicator data encompasses at least five stages and are approximately normal (El-Basyouny, El-Bassiouni 2013; Hoyle 2014).

3. Results of Analysis and Modelling Process

3.1. Estimation Methods

When dealing with the factor analysis and SEM, the studying of normality conditions of observed data is one of the essential issues in the descriptive statistics. The reason is the disturbed accuracy of model validation related statistical tests if the data are severely non-normal (Weston, Gore 2006).

The investigation of non-normality is usually carried-out by the calculation of the Skewness Index (SI) and Kurtosis Index (KI) of the data. There exists certain disagreement in researchers' opinion about the most suitable criterions for the non-normality, which is still acceptable for an effective usage of the ML estimator. However, in general, the researchers agree that the values $|SI| < 3$ and $|KI| < 7$ are still allowable to apply a ML method without any serious concerns (Zhai *et al.* 2013; Lei, Lomax 2005; Weston, Gore 2006; Kline 2015; Ullman 2006).

In the case of our data, the normality conditions were not severely violated, but only slightly. The latter means that the skewness indices of corresponding ordinal indicators were positioned inside the interval $(-1.998, 1.852)$, while their kurtoses indices have taken the values inside the interval $(-1.551, 5.14)$. Therefore, we decided to use the ML estimator since the latter offers quite big spectra of different GOF indices needed in the model validation process.

3.2. Exploratory Factor Analysis

As mentioned, the latest research about the 43-item MRBQ questionnaire had implicated that the subsequent factor model can be articulated by four essential factors, which are speed violations, errors, safety equipment, and stunts (Sakashita *et al.* 2014). Moreover, since

we know that the Q1 questionnaire additionally brings three groups of items ($A_i, i = 1, \dots, 4; C_i, i = 1, \dots, 4; H_i, i = 1, \dots, 3$), we justifiably expect that the EFA will clearly give us the seven-factor solution. By other words, it is anticipated that the three additional factors related to the Q1 (alcohol, helmet, and clothing) will occur separately from the four MRBQ factors, so the seven-factor solution will be the most likely outcome of the EFA.

The possibility that the factor analysis may be used without any concerns was verified by the application of two tests: Kaiser–Meyer–Olkin (KMO) test and the Bartlett's Test of Sphericity (BTS) (Hoyle 2014; Kline 2015; Hair *et al.* 2009). The BTS value was noticeably significant ($\chi^2 = 2777.971$ with $df = 465$ and $p < 0.001$), while the KMO value was $0.743 > 0.5$. According to the recommendation of some authors (Li *et al.* 2013; Frohlich, Westbrook 2001; Sahin *et al.* 2013), the achieved BTS and KMO values imply that the EFA can be reliably conducted in the further research.

In the factors' extraction process, the Principal Axis Factoring (PAF) algorithm, with additional Promax rotation (and Kaiser normalization) was executed. For choosing the optimal number of extracted factors, three criterions were used: the Cattell's scree plot, the Kaiser's 'eigen value bigger than one rule', and the calculation of percentage of variation (Hoyle 2014; Kline 2015). Naturally, only those items were retained in the model, which are significantly loaded on corresponding factors (which means: loadings $\lambda_{ij} > 0.40$, according to Hair *et al.* (2009)).

The results of the rotated factor pattern matrix (loadings, Cronbach's alphas (CAs) and the percent of the total variance explained) are presented in Table 4. Since the CAs of all revealed factors are bigger than the value 0.7, the reliability and internal consistency are adequate (Hair *et al.* 2009). Cumulative percent of the total variance explained (61.017%) is slightly low since 23 ill-fitting items were dropped because of bad communalities and/or inappropriate loadings' properties. However, according to Hair *et al.* (2009), this is not a big problem since in the social sciences the total variance explained is often as low as 50–60% (Williams *et al.* 2010).

As expected, from Table 4 is clear that the alcohol-related items A_i are significantly loaded on the factor 'alcohol' and the clothing-related items C_i are significantly loaded on the factor 'clothing', while the helmet related-items H_i are significantly loaded on the factor 'helmet'. In addition to that, the MRBQ related items M_i are significantly on the corresponding MRBQ factors ('errors', 'speed violations', 'stunts' and 'safety equipment'). Thus, we indeed achieved the seven-factor model solution, as it was a priori assumed. This structure based on seven factors will now represent a useful theoretical guideline for the subsequent CFA.

3.3. Confirmatory Factor Analysis

As mentioned in earlier sections, the CFA is the first stage of creating the structural equation model (Fig. 2). When conducting the CFA, the structure of the previously derived factor model from the EFA was applied as a baseline.

Table 4. Achieved results of the rotated factor pattern matrix (EFA – Exploratory Factor Analysis)

Pattern matrix	Factors						
	Safety equipment	Errors	Stunts	Helmet	Clothing	Speed violations	Alcohol
Cronbach alpha	0.808	0.733	0.785	0.850	0.799	0.790	0.818
% of variance	18.053	11.270	8.162	7.090	6.653	5.387	4.401
Cumulative %	18.053	29.324	37.486	44.575	51.228	56.615	61.017
Items	Item loadings on factors						
M_{30}	0.884						
M_{31}	0.801						
M_{29}	0.751						
M_{32}	0.575						
M_{40}	0.565						
M_3		0.694					
M_1		0.601					
M_5		0.583					
M_{16}		0.518					
M_{11}		0.485					
M_7		0.461					
M_2		0.444					
M_6		0.434					
M_{25}			0.739				
M_{27}			0.680				
M_{28}			0.642				
M_{23}			0.620				
M_{26}			0.540				
H_2				0.970			
H_1				0.885			
H_3				0.635			
C_2					0.964		
C_3					0.772		
C_4					0.587		
M_{19}						0.887	
M_{18}						0.740	
M_{20}						0.543	
M_{21}						0.427	
A_4							0.941
A_3							0.841
A_2							0.471

The CFA procedure checked if this structure is consistent with the measurement theory. To do so the ML method was used for the estimation of all model parameters at first. While estimating the parameters, the difference between the data-based covariance matrix, and the model implied covariance matrix was minimized (Hoyle 2014).

Afterward, the fitting adequacy of the estimated CFA factor model was examined through the computation of several fit indices, which are usually suggested in the scholarly literature (Kline 2015; Byrne 2009; Hoyle 2014; Hair *et al.* 2009). Some of the most typical indices are, for instance (Byrne 2009): Normed Fit Index (NFI), Tucker–Lewis Index (TLI), Comparative Fit Index (CFI), Incremental Fit Index (IFI), Root Mean Square Error of Approximation (RMSEA), Standardized Root Mean Residual (SRMR) and so. As it turned out, all of these indices have taken the adequate value with respect to their

threshold ranges, which are recommended in the literature (Hooper *et al.* 2008; Hair *et al.* 2009; Kline 2015; Hoyle 2014).

To complete the CFA, the derived model had to be evaluated for its convergent and discriminant validity as well. The main issues of the appropriate convergent validity are the adequate Composite Reliability (CR) and Average Variance Extracted (AVE), respectively. Their threshold levels are 0.70 for CR, and 0.5 for AVE (Hair *et al.* 2009). When the discriminant validity is taken into the consideration, the AVE of each factor should be bigger than the squared correlation $CORR^2$ between this factor and any other factor, or $(\sqrt{AVE} > CORR)$ (Hair *et al.* 2009; Fornell, Larcker 1981). As it turned out, the CR values for all seven factors were inside the interval (0.715, 0.862), while the AVE values for all seven factors were inside the interval (0.504, 0.628). Since also the