

## Reliability analysis of geosynthetic-reinforced steep slopes

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1 ABSTRACT: In slope stability analysis, the high degree of uncertainty associated with design  
2 parameters has led to increasing use of reliability-based approaches as a means of evaluating  
3 the combined effects of such uncertainties on the structure performance. In this study, the  
4 reliability level of geosynthetic-reinforced steep slopes designed according to Eurocode 7  
5 (EC7), without any additional margin of safety, was assessed using the commercially-available  
6 *Slide* 6.0 software based on Monte Carlo simulation. To validate the EC7 partial factor design  
7 method regarding structural reliability, the estimated reliability indexes were compared with the  
8 minimum value recommended by Eurocode 0 (EC0). Additionally, through a probabilistic  
9 sensitivity analysis, the effect of variability in design parameters on slope reliability was  
10 evaluated and discussed. The results have shown that the geosynthetic-reinforced slopes  
11 designed to EC7 specifications exhibit generally an adequate reliability level according to EC0.

12 The soil friction angle and the friction angle of the soil-geosynthetic interface and, secondly, the  
13 surcharge load, were found to be the most significant parameters for the reliability of the  
14 analysed slopes. For typical coefficients of variation of design parameters, the EC7 partial factor  
15 method tends to be conservative in terms of structural reliability. However, in situations of  
16 abnormal high variability, the partial factor methodology may lead to unsafe design, and thus  
17 reliability analyses should be implemented.

18 KEYWORDS: Geosynthetics, reliability analysis, geosynthetic-reinforced slope, Monte Carlo  
19 simulation method, probabilistic sensitivity analysis, Eurocodes, variability

20

## 21 **1 INTRODUCTION**

22 The absolute safety of a structure cannot be guaranteed. Uncertainty about loading and  
23 available resistance, limitations of design methods, use of simplifying assumptions and possible  
24 human errors during construction prevent accurate prediction of the structural behaviour and  
25 make it necessary to establish socially tolerable risk levels. The slope stability problem, in  
26 particular, is commonly associated with various sources of uncertainties, such as geological  
27 details missed in the site investigation phase, estimation of soil properties that are difficult to  
28 quantify (i.e. the spatial variability in the field cannot be accurately reproduced), variation in  
29 pore-water pressure, testing errors and many other important factors, which often cannot be  
30 eliminated by reasonable investigate effort or expenditure (Malkawi et al. 2000).

31 The conventional deterministic slope stability analyses consist of determining the global  
32 factor of safety for trial slip surfaces until the slip surface yielding the lowest factor of safety is  
33 located. These analyses are based on fixed representative values for design parameters,  
34 without explicit consideration of their inherent uncertainty and variability. The target factors of  
35 safety, empirically established (i.e. based on past experience), take into account all the  
36 uncertainties and risks involved in the design process. However, since it is common to use the  
37 same factor of safety value for a given type of application, such as long-term slope stability,  
38 without regard to the degree of uncertainty involved in its calculation, the same value of factor of  
39 safety is often applied to conditions that incorporate widely varying degrees of uncertainty

40 (Duncan 2000). In fact, deterministic approaches suffer from several limitations such as the  
41 impossibility of establishing a direct relationship between the global factor of safety and the level  
42 of reliability of a structure and quantifying the impact of uncertainties in random input variables  
43 on the uncertainties of the model outputs.

44 Most current standards (including Structural Eurocodes) are based on semi-probabilistic  
45 safety concepts, using partial safety factors for actions and resistances. The principle behind  
46 semi-probabilistic safety analyses is that uncertainties are treated right at sources with the  
47 introduction of the “characteristic value” and the “design value” of the parameters, which have a  
48 statistical background. If the calculated design value for the effect of actions is lower than the  
49 calculated design value for the resistance, the design fulfils the ultimate limit state requirements.  
50 These analyses include the concepts of uncertainty and risk since there is a correspondence  
51 between the partial safety factors and the reliability index (or probability of failure). According to  
52 Eurocode 0 (EC0), a design using EC0 with the partial factors given in Annex A1 and  
53 Eurocode 1 (EC1) to Eurocode 9 (EC9) is considered generally to lead to a structure with a  
54 reliability index value greater than 3.8 for a 50-year reference period. Holicky and  
55 Vrouwenvelder (2005) stated that the most important advantage of the partial factor design  
56 method is the possibility of taking into account the uncertainty in individual variables by  
57 calibrating the relevant partial factors and other reliability elements. Cardoso and Fernandes  
58 (2001) highlighted the importance of a consistent and rational procedure for defining  
59 characteristic values for geotechnical parameters since the safety provided by the application  
60 of Eurocode 7 (EC7) depends not only on the partial safety factors specified by the code but  
61 also on the way the characteristic values are selected.

62 Probabilistic approaches and reliability analyses have been increasingly applied in slope  
63 stability assessment, as a powerful way of evaluating the combined effects of uncertainty and  
64 variability associated with soil and reinforcement strength parameters and loadings (e.g.  
65 Christian et al. 1994; Kitch 1994; Chowdhury and Xu 1995; Low and Tang 1997a; Griffiths et al.  
66 2007, 2009, 2010; Cho 2010; Fenton and Griffiths 2010; Kitch et al. 2011, Javankhoshdel and  
67 Bathurst 2014). Within a probabilistic framework, the design parameters are treated as random  
68 variables and consequently, the calculated factor of safety is also regarded as a random

69 variable with a probability distribution. The probability of failure and the reliability index may then  
70 be determined and used as performance indicators. Furthermore, probabilistic sensitivity  
71 analyses may help to evaluate the effect of variability in individual parameters and identify the  
72 most influential variables for the structure reliability. However, it is important to note that  
73 probabilistic modelling may be associated with difficulties due to the lack of available  
74 information, typical in geotechnical engineering (Beer et al. 2013). Thus, the simplest and most  
75 obvious advantage of a probabilistic approach or reliability analysis is to complement a  
76 conventional deterministic analysis by incorporating uncertainties associated with the  
77 performance of the geotechnical structure to be analysed, thereby allowing for an enhanced  
78 assessment of the structure reliability and providing an improved basis for interaction between  
79 engineers and decision-makers (Whitman 2000; Chowdhury et al. 2010).

80 Based on results of reliability analyses, risk assessment is often conducted to help  
81 geotechnical engineers in making informed decisions. From an engineering point of view, the  
82 risk is associated with the exposure of recipients to hazards and may be defined as the product  
83 of the probability of an adverse event and its consequences (Baecher and Christian 2003). In  
84 the context of risk analysis, consequence is the outcome or result of a hazard being realised  
85 and may include injury or loss of life, reconstruction costs, loss of economic activity,  
86 environmental losses, among others (Modarres 2006). The process of risk assessment consists  
87 of making a decision recommendation on whether existing risks are acceptable and present  
88 control risk measures are appropriate and, if not, whether alternative measures are justified or  
89 will be implemented. Therefore, risk assessment not only includes the risk analysis, which  
90 involves the definition of scope, danger identification, estimation of probability of occurrence,  
91 evaluation of the vulnerability of the elements at risk, consequence identification and risk  
92 estimation, but also incorporates the risk evaluation, the stage at which the values and  
93 judgement enter the decision process, by including consideration of the relevance of the  
94 estimated risks and the associated social, economic and environmental consequences (Fell et  
95 al. 2005).

96 This paper examines the structural reliability of nine geosynthetic-reinforced steep slopes  
97 designed according to EC7 (CEN 2004), without any additional margin of safety. The probability

98 of failure and the reliability index are estimated using *Slide* 6.0 software (Rocscience Inc. 2010)  
99 based on Monte Carlo simulation. To validate the EC7 partial factor method regarding structural  
100 reliability, the obtained reliability levels are compared with the EC0 (CEN 2002) recommended  
101 minimum value. Furthermore, through a probabilistic sensitivity analysis, the impact of variability  
102 in the input random variables on the slope reliability is investigated and the most significant  
103 design parameters are identified. This paper extends previous work on probabilistic slope  
104 stability analysis presented in Ferreira et al. (2013).

105

## 106 **2 RELIABILITY AND DESIGN UNDER UNCERTAINTY**

### 107 **2.1 The concept of structural reliability**

108 EC0 (CEN 2002) defines reliability as “the ability of a structure or a structural element to  
109 fulfil the specified requirements, including the design working life, for which it has been  
110 designed; reliability is usually expressed in probabilistic terms”. Accordingly, reliability includes  
111 safety, serviceability and durability of a structure.

112 As a measure of reliability, EC0 introduces the reliability index, which may be related to a  
113 probability of failure. In this context, “failure” includes not only catastrophic failure – as in the  
114 case of a landslide – but also any unacceptable difference between expected and observed  
115 performance (Leonards 1975; Baecher and Christian 2003). The relationship between the  
116 reliability index ( $\beta$ ) and the probability of failure (PF) can be expressed as (EC0):

$$PF = \Phi(-\beta) \quad (1)$$

117 where  $\Phi$  is the cumulative distribution function of the standardised normal distribution. Table 1  
118 shows the relationship defined by Equation 1.

119 According to EC0, the probability of failure can be expressed through a performance  
120 function  $g$  such that a structure is considered to survive if  $g > 0$  and fail if  $g < 0$ . Consequently, if  
121  $g$  is normally distributed, the reliability index can be calculated as follows:

$$\beta = \frac{\mu_g}{\sigma_g} \quad (2)$$

122 where  $\mu_g$  is the mean value of  $g$  and  $\sigma_g$  is the respective standard deviation. It is also noted that  
123 in this context,  $\beta$  and PF are only notational values that do not necessarily represent the actual  
124 failure rates but are used as operational values for code calibration purposes and for  
125 comparison of reliability levels of structures.

126 The required level of reliability for a certain structure depends on the consequences that  
127 may arise from a hypothetical failure scenario. EC0 establishes three different Consequences  
128 Classes (CC) based on the potential damage in terms of loss of human life and social,  
129 economic or environmental impact (Table 2). For example, agricultural buildings where people  
130 do not normally enter are comprised in the CC1 class. Residential and office buildings are  
131 included in the CC2 class and grandstands or public buildings, where consequences of failure  
132 are high, integrate the CC3 class.

133 The Consequences Classes (CC1, CC2 and CC3) may be associated with the respective  
134 Reliability Classes (RC1, RC2 and RC3). For each Reliability Class, EC0 establishes  
135 recommended minimum values for the reliability index, as a function of the reference period  
136 (Table 3).

137

## 138 **2.2 Variability and uncertainty in geotechnical design**

139 Geotechnical variability is a complex attribute that results from many sources of  
140 uncertainties. There are three primary sources of geotechnical uncertainties: inherent variability,  
141 measurement errors and transformation uncertainties. Inherent variability arises mainly from the  
142 natural geologic processes that continually modify the in situ soil mass. Measurement errors are  
143 caused by equipment, procedure and/or operator and random testing effects. Equipment effects  
144 result from inaccuracies in the measuring devices and variations in equipment geometries and  
145 systems employed. Procedure and/or operator effects derive from the limitations in existing test  
146 standards and how they are followed. Random testing errors refer to the remaining scatter in  
147 the test results which is neither assignable to specific testing parameters nor caused by inherent  
148 soil variability. The third source of uncertainties (transformation uncertainties) is introduced

149 when field or laboratory measurements are transformed into design properties using correlation  
150 models (Phoon and Kulhawy 1999; Phoon 2004).

151 In the context of reliability analysis, knowledge of distribution type and characterisation of  
152 variability in design parameters are important issues. Normal and lognormal distributions have  
153 been often used in geotechnical design to characterise variability in factor of safety (Duncan  
154 2000; Koerner 2002; Sabatini et al. 2002), permeability, friction angle and unit weight of soil  
155 (Hoeg and Murarka 1974; Lacasse and Nadim 1996; Low and Tang 1997a, 1997b; Phoon and  
156 Kulhawy 1999; Chalermyanont and Benson 2004) and tensile strength of reinforcement (Low  
157 and Tang 1997a; Chalermyanont and Benson 2004). Table 4 indicates typical coefficients of  
158 variation (COV) for design parameters of particular interest for the current study, compiled on  
159 the basis of published data. Typical coefficients of variation for a broad variety of other soil  
160 properties may also be found in the literature (e.g. Lee et al. 1983; Phoon et al. 1995; Lacasse  
161 and Nadim 1996; Baecher and Christian 2003).

162

### 163 **2.3 Monte Carlo simulation method**

164 Several probabilistic methodologies are available for reliability-based design, namely the  
165 Monte Carlo simulation, First Order Second Moment method, Second Order Second Moment  
166 method, Point Estimate method, Hasofer-Lind approach (FORM), among others. Each method  
167 involves different computational effort, provides a different level of accuracy and yields a  
168 different insight into the effects of the individual parameters (Baecher and Christian 2003).

169 The Monte Carlo method provides approximate solutions to a variety of mathematical  
170 problems by performing statistical sampling experiments. The method uses randomly generated  
171 values for the component variables to determine the probability distribution of the design  
172 variable (e.g. factor of safety). Its application requires the knowledge of the statistical  
173 distribution of the input random variables. The steps for the implementation of the Monte Carlo  
174 method may be outlined as follows (Dai et al. 1993).

175 1. Generation of random numbers which are independent random variables uniformly  
176 distributed over the unit interval between zero and one.

- 177 2. Transformation of the random numbers from a uniform distribution to the distribution  
178 applicable to the component variable.
- 179 3. Calculation of values of all component variables based on the appropriate random  
180 numbers.
- 181 4. Computation of the design variable (e.g. factor of safety) using the generated values of  
182 the component variables.
- 183 5. Repetition of steps 1. to 4. for a large number of times. The number of times these steps  
184 are repeated depends on the variability of the input and output parameters and the  
185 desired accuracy of the output.
- 186 6. Creation of a cumulative distribution of the design function using the data obtained from  
187 the above simulations.

188 The method is conceptually simple and has the capability of dealing with a wide range of  
189 functions, even those that cannot be expressed conveniently in explicit form. However, it has  
190 the disadvantage that it may converge slowly. Further details of this approach have been  
191 presented over recent decades by several authors, namely Hammersley and Handscomb  
192 (1964), Schreider (1966), Rubinstein (1981), Fishman (1996) and Baecher and Christian (2003).

193

#### 194 **2.4 Probabilistic sensitivity analysis**

195 Sensitivity analyses have been widely applied in different areas of science and technology,  
196 such as engineering design, to investigate how a given model output depends upon the input  
197 parameters. This can be motivated simply by the wish of understanding the implications of a  
198 complex model but often arises due to the uncertainty about the true values that should be used  
199 for the input parameters (Oakley and O'Hagan 2004).

200 Among the different methods of sensitivity analysis, probabilistic sensitivity analysis is  
201 generally considered to be the most rigorous and is gaining widespread use. In design under  
202 uncertainty, probabilistic sensitivity analyses are typically performed to quantify the impact of  
203 uncertainties in random input variables (characterised by a probability distribution) on the

204 uncertainty of the model output. Results from probabilistic sensitivity analyses have been used  
205 in engineering design for a range of purposes, including (Saltelli et al. 2000):

- 206 – reducing the dimension of a design problem by identifying the probabilistically  
207 insignificant factors;
- 208 – checking the validity of a model and the assumptions made on the probability  
209 distributions of the random inputs;
- 210 – obtaining insights into the design space and the probabilistic behaviour of a model  
211 response;
- 212 – investigating potential improvement on the probabilistic response by reducing the  
213 uncertainty in random inputs.

214 When applied to risk assessment, probabilistic sensitivity analyses can be very useful for  
215 understanding how risk estimates and, particularly, risk-based decisions are dependent on the  
216 variability and uncertainty in factors contributing to risk. In other words, sensitivity analyses can  
217 help to identify what is governing the risk estimates and, in these circumstances, contribute to  
218 risk mitigation by reducing the uncertainty related to the most relevant variables. This may be  
219 accomplished, for instance, by means of complementary geotechnical investigation (e.g. field  
220 investigation, laboratory testing, etc.).

221

### 222 **3 GEOSYNTHETIC-REINFORCED SLOPE MODELS**

223 In this study, nine geosynthetic-reinforced steep slopes designed according to EC7  
224 (CEN 2004), without any additional margin of safety, were modelled and analysed using  
225 *Slide 6.0* software (Rocscience Inc. 2010). The geometry of the reinforced slopes is shown in  
226 Figure 1. All nine slopes had height  $H = 8.4$  m and were assumed to rest on competent  
227 foundations. The reinforcement layout consisted of fourteen horizontal geogrid layers with  
228 constant length ( $L$ ) and vertical spacing  $s = 0.6$  m throughout the slope. For each slope, two  
229 geogrids with different tensile strengths were considered so that a stronger geogrid (GGR1) was  
230 used for the seven lower layers and a weaker geogrid (GGR2) was employed near the top of

231 the structure. The backfill material was assumed to be a cohesionless granular soil with design  
232 unit weight  $\gamma_d = 23 \text{ kN/m}^3$ .

233 Three different slope angles were considered:  $\alpha = 60^\circ$  (Slopes 1 to 7),  $\alpha = 45^\circ$  (Slope 8) and  
234  $\alpha = 75^\circ$  (Slope 9). For Slopes 1 to 4, the design values of the soil internal friction angle ( $\phi_d$ ) were  
235 taken as  $20^\circ$ ,  $25^\circ$ ,  $30^\circ$  and  $35^\circ$ , respectively, and no surcharge load was imposed. In the case of  
236 Slopes 5 to 7, the design friction angles of the soil were respectively equal to  $25^\circ$ ,  $30^\circ$  and  $35^\circ$ ,  
237 but a uniform vertical surcharge ( $S_d$ ) of 13 kPa (design value) was applied on the slope crest.  
238 Slopes 8 and 9 were also subjected to a surcharge load  $S_d = 13 \text{ kPa}$  and the design friction  
239 angle of the backfill material was set at  $25^\circ$ . The soil-geogrid interface friction angle ( $\delta_d$ ) was  
240 defined in terms of a  $\delta_d/\phi_d$  ratio ( $\delta_d/\phi_d = 6/7$ ) which was held constant for all slopes and both  
241 geogrid reinforcements (GGR1 and GGR2).

242 The design tensile strength ( $T_d$ ) of geogrids GGR1 and GGR2 and the reinforcement length  
243 provided were checked, following the design procedure proposed by Jewell (1989, 1996), so  
244 that the internal and overall equilibrium of the slopes was satisfied. It should be noted that  
245 Jewell's charts apply to reinforced slopes with a level crest and resting on a competent  
246 foundation, which is the case of the slopes herein investigated.

247 As previously mentioned, Structural Eurocodes adopt a semi-probabilistic approach for  
248 safety verification, using design values for actions and resistances. In a common design  
249 process, the design values of the variables would be determined from their characteristic values  
250 using partial safety factors. However, in the present study, the design values were defined first,  
251 to ensure that the design load was equal to the design strength. The characteristic values were  
252 then back-calculated, using partial safety factors in accordance with the Combination 2 of the  
253 Design Approach 1 of EC7 for verification of the ultimate limit state GEO (related to failure or  
254 excessive deformation of the ground) in persistent and transient situations. Since the code does  
255 not specify which partial safety factor should be used for the determination of the design tensile  
256 strength of geosynthetics, a partial safety factor meeting the requirements of the ISO/TR  
257 20432:2007 (ISO 2007) was considered.

258 The mean values of the design parameters to be used in the probabilistic stability analysis  
259 of the reinforced slopes were then determined, assigning a statistical distribution to the design  
260 parameters which were considered as random variables (soil unit weight, soil friction angle, soil-  
261 geogrid interface friction angle, tensile strength of geogrids and surcharge load). With the  
262 exception of the surcharge load, the variables were assumed to be normally distributed. Typical  
263 coefficients of variation were assigned to each of the distributions using data reported in the  
264 literature (see Table 4). For these variables, the characteristic values were assumed as  
265 quantiles of 5% or 95% of the statistical distributions depending on whether the parameters  
266 contribute to safety or not, respectively. The surcharge load was statistically characterised by an  
267 exponential distribution and its mean value was set equal to its characteristic value.

268 Table 5 presents the partial safety factors (PSF) and the coefficients of variation (COV)  
269 adopted in this study. Table 6 lists the design values (DV), the characteristic values (CHV) and  
270 the mean values (MV) of the design parameters for all the slopes analysed.

271 Using the simplified Bishop's method, which is one of the most commonly adopted limit  
272 equilibrium methods for slope stability analysis and is widely accepted as reasonably accurate,  
273 the deterministic Global Minimum circular slip surfaces and the ratio of the design strength to  
274 the design effect of actions (the so-called over-design factor) were obtained (Figures 2 to 10).  
275 Since the over-design factors are equal to unity, Slopes 1 to 9 fulfil the EC7 safety  
276 requirements, but no additional margin of safety is established.

277

#### 278 **4 RELIABILITY ANALYSIS OF GEOSYNTHETIC-REINFORCED SLOPES**

279 There are two types of probabilistic stability analysis which may be carried out with *Slide 6.0*  
280 software (Rocscience Inc. 2010): the Global Minimum method (fixed method) and the Overall  
281 Slope method (floating method). With the Global Minimum option, the probabilistic analysis is  
282 carried out only on the Global Minimum slip surface located by the deterministic slope stability  
283 analysis. The factor of safety of this single surface is recalculated  $n$  times (where  $n$  is the  
284 number of Monte Carlo simulations) using random values for the input parameters. The  
285 probability of failure is then computed as the number of analyses which result in a factor of  
286 safety less than unity, divided by the total number of samples. With this approach, the

287 probability of failure (or the reliability index) of the deterministic Global Minimum slip surface is  
288 considered representative of the probability of failure for the slope.

289 With the Overall Slope analysis type, the entire search for a Global Minimum slip surface is  
290 repeated  $n$  times. For each search iteration, a new set of random variables is first determined  
291 and the Global Minimum slip surface is then located. The Overall Slope reliability is based on  
292 the distribution of factors of safety obtained for all the Global Minimum slip surfaces located by  
293 the analysis. Since several Global Minimum slip surfaces are generally encountered, the  
294 probability of failure and the reliability index calculated for the overall slope are not associated  
295 with a specific slip surface. A potential advantage of the Overall Slope method, when compared  
296 with the Global Minimum method, is that it does not assume that the probability of failure of the  
297 slope is equal to the probability of failure of the deterministic Global Minimum slip surface.  
298 However, it involves a substantially greater computation time.

299 The application of the Monte Carlo simulation method requires that design parameters be  
300 characterised by their probability distributions, which describe the range of possible input values  
301 along with their probability of occurrence. As mentioned before, the unit weight ( $\gamma$ ) and friction  
302 angle ( $\phi$ ) of the soil, the soil-geogrid interface friction angle ( $\delta$ ), the tensile strength of the  
303 geogrids ( $T$ ) and the surcharge load ( $S$ ) were treated as random variables. The coefficients of  
304 variation for the input variables which were assumed to follow normal distributions ( $\gamma$ ,  $\phi$ ,  $\delta$  and  $T$ )  
305 were previously indicated in Table 5. The mean values for all the design parameters were  
306 presented in Table 6. The surcharge load considered in the design of Slopes 5 to 9 was  
307 assigned an exponential distribution defined by a mean of 10 kPa and minimum and maximum  
308 values respectively equal to 0 kPa and 30 kPa, aiming to cover the high degree of uncertainty  
309 often associated with this variable.

310 Figure 11 illustrates the evolution of the reliability index of one example slope (Slope 5) as a  
311 function of the number of Monte Carlo simulations, obtained from the Global Minimum and  
312 Overall Slope methods. The number of random trials adopted ( $n = 1\,000\,000$ ) was high enough  
313 to ensure the convergence of the simulations, and hence adequate accuracy in the results. This  
314 number of Monte Carlo simulations was maintained in all the analyses performed.

315 Figures 12 and 13 compare additional data from the probabilistic stability analysis of  
316 Slope 5, carried out using the Global Minimum and Overall Slope options. The probability  
317 density functions and the cumulative probability distributions of the factor of safety determined  
318 from the former methods are presented in Figure 12. Figure 13 shows the values of the mean  
319 factor of safety, probability of failure and reliability index (assuming a normal distribution of the  
320 factor of safety results). The reliability indexes estimated from the Global Minimum and Overall  
321 Slope methods were respectively 4.141 (Figure 13a) and 4.130 (Figure 13b), which exceed the  
322 target value established by EC0 for structures of RC2 and a 50-year reference period ( $\beta = 3.8$ ).  
323 Also visible in Figure 13b are the multiple Global Minimum slip surfaces which were located  
324 throughout the Overall Slope probabilistic analysis. Nevertheless, the difference between the  
325 results obtained from both methods is not significant (Figures 12 and 13), which may be  
326 attributed to the fact that the correlation between the factors of safety of different failure  
327 surfaces is very high. As pointed out by Cornell (1967), for the case of highly correlated failure  
328 modes, the contribution to the system probability of failure from failure surfaces other than that  
329 associated with the maximum probability of failure may be small, even though they are  
330 numerous. However, it has been mentioned in the literature that for cohesive soil slopes with  
331 spatial variability in the soil parameters, the overall probability of failure may be significantly  
332 higher than the probability of failure associated with a fixed critical slip surface (Cho 2010;  
333 Javankhoshdel and Bathurst 2014).

334 The results from the reliability analysis of the different slopes investigated in the current  
335 study are summarised in Table 7. Regardless of the reinforced slope considered, the different  
336 analysis types (Global Minimum and Overall Slope) provided quite similar results. It can be  
337 observed that the values of the factor of safety and the reliability index increased progressively  
338 with the slope angle (Slopes 8, 5 and 9), which is in agreement with the results reported by  
339 Kitch (1994) for two geogrid-reinforced slopes with slope angles of 45° and 70°. The data  
340 presented in Table 7 also show that the reliability index decreased progressively as the friction  
341 angle of the soil was increased (Slopes 1 to 4 and Slopes 5 to 7). In other words, the reliability  
342 index was found to decrease as the relative contribution of the soil shear strength to the slope  
343 stability increased (i.e. when the soil friction angle increased, and hence the required

344 reinforcement contribution to strength decreased) and, on the other hand, it was found to  
345 increase when the relative contribution of the reinforcement tensile strength to the slope stability  
346 increased (i.e. when the slopes became steeper and the tensile strength of the geogrids was  
347 increased so as to ensure the slope stability). This may be justified by the fact that the  
348 coefficient of variation of the soil friction angle (COV = 7%) was higher than that of the geogrid  
349 tensile strength (COV = 5%), and thus decreasing the uncertainty associated with the strength  
350 properties increases the reliability index and vice-versa.

351 For the number of simulations performed ( $n = 1\,000\,000$ ), all the factors of safety obtained  
352 corresponded to safe situations ( $FS \geq 1$ ), and hence the computed probabilities of failure for  
353 Slopes 1 to 9 were equal to zero (Table 7). In any case, even considering the normal  
354 distribution fit for this zone, the number of points in the vicinity of  $FS = 1$  was limited. To more  
355 accurately characterise the probability of failure using the Monte Carlo method, the use of a  
356 technique able to generate a relevant number of sets of values for design parameters resulting  
357 in a factor of safety close to unity would be required.

358 As shown in Table 7, with the exception of Slopes 4 and 7, whose soil friction angle was the  
359 highest considered in this study ( $\phi_d = 35^\circ$ ), the reliability indexes for the different geogrid-  
360 reinforced slopes were greater than the EC0 recommended minimum value ( $\beta = 3.8$ ). This  
361 finding supports the idea that, for common structures, the EC7 partial factor design method  
362 tends to be conservative from the point of view of structural reliability.

363

## 364 **5 PROBABILISTIC SENSITIVITY ANALYSIS**

### 365 **5.1 General**

366 In design under uncertainty, probabilistic sensitivity analyses are commonly performed to  
367 evaluate the effect of variability in random input parameters on the probabilistic characteristics  
368 of a design performance. Results from probabilistic sensitivity analyses may be particularly  
369 useful when used as a design aid in decision making for selecting the most suitable design  
370 based on specific project constraints (e.g. target cost or schedule).

371 In the present study, *Slide* 6.0 software was used to perform a probabilistic sensitivity  
372 analysis of the variability associated with the design parameters of Slopes 5, 6, 8 and 9. The  
373 primary objectives of this analysis were the following: to ascertain how the variability in design  
374 parameters influences the reliability level of geosynthetic-reinforced steep slopes with different  
375 slope angles and soil friction angles; to identify the most relevant parameters regarding the  
376 reliability of these structures; and to assess the level of safety margin of the EC7 partial factor  
377 method with respect to structural reliability, which may be of particular importance in the  
378 absence of clear information concerning the variability associated with design parameters.

379 To perform the sensitivity analysis, different COVs around the most likely value (the value  
380 adopted in the reliability analysis presented in the previous section) were assigned to soil and  
381 reinforcement parameters ( $\gamma$ ,  $\phi$ ,  $\delta$  and  $T$ ) on the basis of published data (see Table 4). With  
382 respect to the surcharge load ( $S$ ), the analysis was carried out by varying the upper bound  
383 value. Table 8 indicates the COVs and surcharge limits considered. For each soil or  
384 reinforcement parameter, its COV was varied within the considered range, while the COVs  
385 corresponding to the remaining parameters and the surcharge limits were held constant at their  
386 most likely values. Similarly, for the probabilistic sensitivity analysis of the variability associated  
387 with the surcharge load, the COVs of the soil and reinforcement parameters were kept constant  
388 and equal to their most likely values. For each combination of COVs and surcharge limits, a  
389 probabilistic stability analysis (using the Global Minimum option) with 1 000 000 Monte Carlo  
390 simulations was performed and the model response was evaluated. Given the similarity  
391 between the results obtained from the Global Minimum and Overall Slope methods in the  
392 previous probabilistic analyses (see section 4), the added computation time required to carry  
393 out the sensitivity analyses using the Overall Slope option seemed unwarranted.

394

## 395 **5.2 COV of soil unit weight**

396 Table 9 presents the results of the probabilistic sensitivity analysis carried out to investigate  
397 how the uncertainty in the soil unit weight may affect the reliability of Slopes 5, 6, 8 and 9, in  
398 terms of the mean factor of safety, probability of failure and reliability index.

399 From Table 9 it can be concluded that the variation of the COV of the soil unit weight from  
400 1% to 10% did not significantly affect the reliability level of the reinforced slopes, which could be  
401 expected since the soil weight influences both the normal and shear forces acting on each slice.  
402 In fact, although a slight reduction in the reliability index may be identified, the values of the  
403 mean factor of safety obtained for each slope remained nearly constant as the COV of the soil  
404 unit weight was increased. Furthermore, it is possible to observe that all the calculated reliability  
405 indexes were above the EC0 reference value for a structure of RC2 and a 50-year reference  
406 period ( $\beta = 3.8$ ). These results suggest that, for conditions similar to those adopted in this study,  
407 a reinforced slope designed according to the partial factor method proposed by EC7 maintains  
408 an adequate level of reliability even if the real COV of the soil unit weight reaches 10%,  
409 provided that the variability in the remaining design parameters corresponds to the expected  
410 value (the value considered in the slope design).

411

### 412 **5.3 COV of soil friction angle and soil-geogrid interface friction angle**

413 The influence of the simultaneous variation of the COV of the soil friction angle and soil-  
414 geogrid interface friction angle was evaluated by using different COVs ranging from 2% to 15%.  
415 The results from these simulations are shown in Table 10. The probability of failure and the  
416 reliability index for all four slopes were significantly affected by the increase in the variability  
417 associated with the shear strength of the soil and soil-geogrid interface. When the COV of the  
418 friction angles was increased up to 10% or 15%, the values of the reliability index fell below the  
419 EC0 reference value and the probability of failure reached maximum values of about 1.2%,  
420 1.6%, 2.1% and 0.4% for Slopes 5, 6, 8 and 9, respectively. These results suggest that the  
421 friction angles of soil and soil-geogrid interface play a decisive role on the slope reliability,  
422 regardless of the slope angle. Kitch (1994) and Kitch et al. (2011) obtained a similar conclusion  
423 regarding the effect of the variability in the soil shear strength on the reliability of geogrid-  
424 reinforced slopes, using the first-order reliability method (FORM). Thus, in the design and  
425 stability analysis of geosynthetic-reinforced steep slopes, efforts should be made in order to  
426 properly characterise the uncertainty associated with the soil and soil-geosynthetic interface

427 shear strength, which may be accomplished by means of direct shear tests on soil and soil-  
428 geosynthetic pullout and direct shear tests (e.g. Sukmak et al. 2015; Hatami and Esmaili, 2015;  
429 Ferreira et al. 2015a, 2015b), using the specific materials to be used on the project. The  
430 obtained results also suggest that a reinforced slope designed according to EC7, without any  
431 additional margin of safety, will probably not be a reliable structure (according to EC0) if the real  
432 COVs of the soil and interface friction angles reach values higher than those adopted in the  
433 slope design.

434

#### 435 **5.4 COV of geogrid tensile strength**

436 In order to understand how the variability in the geogrid tensile strength may affect the  
437 reliability of geogrid-reinforced steep slopes designed according to EC7, different COVs in the  
438 range of 1% to 10% were assigned to this variable (Table 11). The results presented in  
439 Table 11 demonstrate that the variation of the COV of the geogrid tensile strength over the  
440 considered range did not have a relevant influence on the mean factor of safety, probability of  
441 failure and reliability index of these particular slopes. Despite being of little importance, the  
442 influence of the variability in the geogrid tensile strength on the obtained reliability indexes  
443 increased with the slope angle (Slopes 8, 5 and 9), which is consistent with the results  
444 presented by Kitch (1994) and Kitch et al. (2011). Regardless of the variability in the geogrid  
445 tensile strength, the values of the reliability index were greater than the EC0 recommended  
446 value for structures of RC2 and a reference period of 50 years ( $\beta = 3.8$ ). Therefore, the geogrid  
447 tensile strength is not considered to be a significant design parameter with respect to the  
448 structural reliability of the analysed slopes. However, this is probably related to the fact that the  
449 mobilised tensile strength of the geogrids is well below their design strength. In cases where the  
450 previous values are closer, the variability in the geogrid tensile strength may have much more  
451 impact on slope reliability. As shown by Kitch (1994) and Kitch et al. (2011), for internal failure  
452 modes (critical slip surfaces passing predominantly through the reinforced portion of the slope),  
453 the variability in the geogrid tensile strength may have a marked influence on the reliability level  
454 of geogrid-reinforced slopes.

## 455        **5.5    Upper limit of the surcharge load**

456        The effect of the variability in a surcharge load (with a mean value of 10 kN/m<sup>2</sup>) applied on  
457        the top of the slopes was studied by defining different upper bound values for the corresponding  
458        exponential probability distribution (from 15 kPa to 50 kPa). The results from this sensitivity  
459        analysis are presented in Table 12. As can be noted from the table, the increase in the  
460        variability of the surcharge load induced a small reduction in the reliability index of the slopes.  
461        Moreover, it is important to highlight that for Slopes 6 and 8, the reliability index obtained when  
462        the upper limit of the surcharge load was set at 50 kPa was slightly lower than the EC0  
463        recommended minimum value ( $\beta = 3.8$ ).

464

## 465        **5.6    Limit combinations**

466        In order to better understand to what extent the combined variability in the design  
467        parameters may affect the reliability of geosynthetic-reinforced slopes, two additional  
468        combinations were analysed in which the variability in all the input parameters was set very low  
469        or abnormally high (Table 13).

470        From Table 13 it can be concluded that if the variability associated with all the design  
471        parameters is very low, the reliability index of each slope more than triplicates the value  
472        obtained from the probabilistic analysis in which the most likely COVs and surcharge limits were  
473        considered. In contrast, as would be expected, if the variability is abnormally high, the  
474        probability of failure substantially increases and the reliability index undergoes a sharp  
475        reduction. As shown in Table 13, the minimum reliability index obtained for the studied slopes  
476        was about 1.8, which is less than half the value recommended by EC0 for structures of RC2  
477        and a reference period of 50 years, corresponding to totally unacceptable probabilities of failure.

478

## 479        **6        CONCLUSIONS**

480        This paper investigates the structural reliability of geosynthetic-reinforced steep slopes  
481        designed according to EC7 (without any additional margin of safety), using the Monte Carlo  
482        method, and compares the estimated levels of reliability with the EC0 recommended minimum

483 value. A probabilistic sensitivity analysis is then performed, enabling the evaluation of the effect  
484 of the variability associated with input random variables (i.e. soil and reinforcement parameters  
485 and loadings) on slope reliability. Based on the obtained results, the following conclusions can  
486 be drawn.

487 Among the nine geosynthetic-reinforced steep slopes analysed in this study, seven  
488 exhibited a reliability index greater than the EC0 recommended minimum value for structures of  
489 Reliability Class 2 and a reference period of 50 years ( $\beta = 3.8$ ). Only those whose soil friction  
490 angle was the highest herein considered (design value of  $35^\circ$ ) presented a reliability index  
491 slightly lower than the EC0 target value. Therefore, for usual values of design parameters, the  
492 EC7 partial factor method leads generally to a structure meeting the EC0 requirements in terms  
493 of structural reliability.

494 Since the variability and uncertainty associated with the reinforcement strength is typically  
495 lower than that related to the ground strength properties, the reliability index of the geosynthetic-  
496 reinforced slopes decreased as the relative contribution of the soil shear strength to the slope  
497 stability increased (i.e. when the soil friction angle increased, and consequently the required  
498 reinforcement contribution to strength decreased) and, on the other hand, it increased when the  
499 relative contribution of the reinforcement strength to the slope stability increased (i.e. when the  
500 slopes became steeper and the tensile strength of the geogrids was increased so as to ensure  
501 the slope stability).

502 No relevant differences between the results of the reliability analyses carried out using fixed  
503 or floating probabilistic methods (i.e. the Global Minimum and Overall Slope methods available  
504 in *Slide* 6.0 software) were observed.

505 The probabilistic sensitivity analysis of the variability associated with the input random  
506 variables revealed that the soil friction angle and that of the soil-geosynthetic interface and,  
507 secondly, the surcharge load, were the design parameters that had the most influence on the  
508 reliability of the investigated slopes. Therefore, in design and stability analysis of geosynthetic-  
509 reinforced slopes, efforts should be directed at reducing the uncertainty associated with such  
510 parameters.

511 For common variability in design parameters, the EC7 partial factor method tends to be  
512 conservative from the point of view of reliability. In any case, for situations where the variability  
513 in the input parameters reaches abnormal high values, the partial factor methodology may lead  
514 to unsafe design, and hence reliability analyses should be implemented.

515

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521

## 522 **NOTATION**

523 Basic SI units are given in parentheses.

524  $H$  – slope height (m)

525  $L$  – reinforcement length (m)

526  $n$  – number of Monte Carlo simulations (dimensionless)

527  $s$  – vertical spacing of reinforcement layers (m)

528  $S$  – surcharge load (N/m<sup>2</sup>)

529  $S_d$  – design value of surcharge load (N/m<sup>2</sup>)

530  $T$  – geogrid tensile strength (N/m)

531  $T_d$  – design value of geogrid tensile strength (N/m)

532  $\alpha$  – slope angle (degrees)

533  $\beta$  - reliability index (dimensionless)

534  $\gamma$  – soil unit weight (N/m<sup>3</sup>)

535  $\gamma_d$  – design value of soil unit weight (N/m<sup>3</sup>)

536  $\delta$  – soil-geosynthetic interface friction angle (degrees)

537  $\delta_d$  – design value of soil-geosynthetic interface friction angle (degrees)

538  $\phi$  – soil friction angle (degrees)

539  $\phi_d$  – design value of soil friction angle (degrees)

540

#### 541 **ABBREVIATIONS**

542 CC – Consequences Class

543 CHV – characteristic value

544 COV – coefficient of variation

545 DV – design value

546 EC0 – Eurocode 0 (CEN 2002)

547 EC7 – Eurocode 7 (CEN 2004)

548 FS – factor of safety

549 GEO – ultimate limit state related to failure or excessive deformation of the ground (EC7)

550 GGR – geogrid

551 MV – mean value

552 PF – probability of failure

553 PSF – partial safety factor

554 RC – Reliability Class

555

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Figure 4. Global Minimum slip surface and over-design factor of Slope 3 ( $\alpha = 60^\circ$ ,  $L = 4.2$  m,  $\phi_d = 30^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

Figure 5. Global Minimum slip surface and over-design factor of Slope 4 ( $\alpha = 60^\circ$ ,  $L = 3.4$  m,  $\phi_d = 35^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

Figure 6. Global Minimum slip surface and over-design factor of Slope 5 ( $\alpha = 60^\circ$ ,  $L = 5.8$  m,  $\phi_d = 25^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

Figure 7. Global Minimum slip surface and over-design factor of Slope 6 ( $\alpha = 60^\circ$ ,  $L = 4.5$  m,  $\phi_d = 30^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

Figure 8. Global Minimum slip surface and over-design factor of Slope 7 ( $\alpha = 60^\circ$ ,  $L = 3.6$  m,  $\phi_d = 35^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

Figure 9. Global Minimum slip surface and over-design factor of Slope 8 ( $\alpha = 45^\circ$ ,  $L = 6.3$  m,  $\phi_d = 25^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

Figure 10. Global Minimum slip surface and over-design factor of Slope 9 ( $\alpha = 75^\circ$ ,  $L = 5.8$  m,  $\phi_d = 25^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

Figure 11. Convergence of the reliability index of Slope 5 (Global Minimum and Overall Slope methods)

Figure 12. Comparison of results of the probabilistic stability analysis of Slope 5 obtained from the Global Minimum and Overall Slope methods: a) probability density function of the factor of safety; b) cumulative probability distribution of the factor of safety

Figure 13. Results of the probabilistic stability analysis of Slope 5: a) Global Minimum method; b) Overall Slope method

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*Reliability analysis of geosynthetic-reinforced steep slopes, Geosynthetics International, Vol. 23, Issue 4, pp. 301-315, <https://doi.org/10.1680/jgein.15.00057>*

# **TABLES**

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Table 1. Relationship between  $\beta$  and PF (modified from EC0; CEN (2002))

PF	$10^{-1}$	$10^{-2}$	$10^{-3}$	$10^{-4}$	$10^{-5}$	$10^{-6}$	$10^{-7}$
$\beta$	1.28	2.32	3.09	3.72	4.27	4.75	5.20

**Table 2. Definition of Consequences Classes (modified from EC0; CEN (2002))**

Consequences Class	Consequence	
	Loss of human life	Social, economic and environmental
CC1	Low	Small/Negligible
CC2	Medium	Considerable
CC3	High	Very great

Table 3. Recommended minimum values for  $\beta$  for ultimate limit states design (modified from EC0; CEN (2002))

Reliability Class	Minimum values for $\beta$	
	1 year reference period	50 years reference period
RC1	4.2	3.3
RC2	4.7	3.8
RC3	5.2	4.3

Table 4. Typical coefficients of variation of design parameters

Parameter	COV (%)	Source
Soil unit weight	3 - 7	Harr (1984), Kulhawy (1992)
Soil friction angle (granular soil)	2 - 15	Singh (1971), Lumb (1974), Hoeg and Murarka (1974), Schultze (1975), Harr (1984), Kulhawy (1992), Phoon et al. (1995)
Soil-geosynthetic interface friction angle <sup>3</sup>	3.7 - 5.4 <sup>1</sup> 10.2 - 16.7 <sup>2</sup>	Sia and Dixon (2007)
Geosynthetic tensile strength <sup>3</sup>	1.4 - 6.8	Silvano (2005), Lopes et al. (2006), Vieira (2008), Morais (2010), Pinho-Lopes and Lopes (2013)

<sup>1</sup>Values corresponding to coarse grained soil-geosynthetic interfaces.

<sup>2</sup>Values corresponding to fine grained soil-geosynthetic interfaces.

<sup>3</sup>Based on repeatability testing programmes.

**Table 5. Partial safety factors and coefficients of variation of the design parameters**

Parameter	PSF	COV (%)
Soil unit weight	1	5
Soil friction angle	1.25 <sup>1</sup>	7
Soil-geogrid interface friction angle	1.25 <sup>1</sup>	7
Long-term tensile strength of the geogrids	1.25	5
Surcharge load	1.3	-

<sup>1</sup>Applied to the tangent of the friction angle.

Table 6. Design values, characteristic values and mean values of the design parameters

	Parameter	DV	CHV	MV
Slope 1 $\alpha = 60^\circ$ $L = 7.2$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	20.0	24.5	27.6
	Soil-geogrid interface friction angle (°)	17.1	21.0	23.8
	Geogrid tensile strength - GGR1 (kN/m)	34.6	43.3	47.1
	Geogrid tensile strength - GGR2 (kN/m)	19.8	24.8	27.0
Slope 2 $\alpha = 60^\circ$ $L = 5.4$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	25.0	30.2	34.2
	Soil-geogrid interface friction angle (°)	21.4	26.1	29.5
	Geogrid tensile strength - GGR1 (kN/m)	24.7	30.9	33.6
	Geogrid tensile strength - GGR2 (kN/m)	14.1	17.6	19.2
Slope 3 $\alpha = 60^\circ$ $L = 4.2$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	30.0	35.8	40.5
	Soil-geogrid interface friction angle (°)	25.7	31.0	35.1
	Geogrid tensile strength - GGR1 (kN/m)	17.3	21.6	23.6
	Geogrid tensile strength - GGR2 (kN/m)	9.9	12.4	13.5
Slope 4 $\alpha = 60^\circ$ $L = 3.4$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	35.0	41.2	46.6
	Soil-geogrid interface friction angle (°)	30.0	35.8	40.5
	Geogrid tensile strength - GGR1 (kN/m)	11.7	14.6	15.9
	Geogrid tensile strength - GGR2 (kN/m)	9.0	11.3	12.3
Slope 5 $\alpha = 60^\circ$ $L = 5.8$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	25.0	30.2	34.2
	Soil-geogrid interface friction angle (°)	21.4	26.1	29.5
	Geogrid tensile strength - GGR1 (kN/m)	26.3	32.9	35.8
	Geogrid tensile strength - GGR2 (kN/m)	15.7	19.6	21.4
	Surcharge load (kPa)	13.0	10.0	10.0
Slope 6 $\alpha = 60^\circ$ $L = 4.5$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	30.0	35.8	40.5
	Soil-geogrid interface friction angle (°)	25.7	31.0	35.1
	Geogrid tensile strength - GGR1 (kN/m)	18.5	23.1	25.2
	Geogrid tensile strength - GGR2 (kN/m)	11.1	13.9	15.1
	Surcharge load (kPa)	13.0	10.0	10.0
Slope 7 $\alpha = 60^\circ$ $L = 3.6$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	35.0	41.2	46.6
	Soil-geogrid interface friction angle (°)	30.0	35.8	40.5
	Geogrid tensile strength - GGR1 (kN/m)	12.5	15.6	17.0
	Geogrid tensile strength - GGR2 (kN/m)	10.0	12.5	13.6
	Surcharge load (kPa)	13.0	10.0	10.0
Slope 8 $\alpha = 45^\circ$ $L = 6.3$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	25.0	30.2	34.2
	Soil-geogrid interface friction angle (°)	21.4	26.1	29.5
	Geogrid tensile strength - GGR1 (kN/m)	17.1	21.4	23.3
	Geogrid tensile strength - GGR2 (kN/m)	10.3	12.9	14.0
	Surcharge load (kPa)	13.0	10.0	10.0
Slope 9 $\alpha = 75^\circ$ $L = 5.8$ m	Soil unit weight (kN/m <sup>3</sup> )	23.0	23.0	21.3
	Soil friction angle (°)	25.0	30.2	34.2
	Soil-geogrid interface friction angle (°)	21.4	26.1	29.5
	Geogrid tensile strength - GGR1 (kN/m)	36.3	45.4	49.4
	Geogrid tensile strength - GGR2 (kN/m)	21.7	27.1	29.6
	Surcharge load (kPa)	13.0	10.0	10.0

**Table 7. Results of the probabilistic stability analysis of Slopes 1 to 9**

	FS (deterministic)	Global Minimum ( $n = 1\,000\,000$ )			Overall Slope ( $n = 1\,000\,000$ )		
		FS (mean)	PF <sup>1</sup> (%)	$\beta$	FS (mean)	PF <sup>1</sup> (%)	$\beta$
Slope 1	1.450	1.453	0	4.344	1.449	0	4.464
Slope 2	1.462	1.466	0	4.228	1.461	0	4.167
Slope 3	1.487	1.492	0	3.827	1.492	0	3.901
Slope 4	1.533	1.542	0	3.631	1.543	0	3.641
Slope 5	1.476	1.489	0	4.141	1.489	0	4.130
Slope 6	1.506	1.519	0	3.838	1.516	0	3.893
Slope 7	1.542	1.559	0	3.618	1.558	0	3.639
Slope 8	1.475	1.487	0	3.852	1.485	0	3.847
Slope 9	1.484	1.497	0	4.587	1.498	0	4.591

<sup>1</sup>Determined as the ratio of the number of simulations with FS < 1 to the total number of simulations ( $n = 1\,000\,000$ ).

**Table 8. Coefficients of variation and surcharge limit values used in the probabilistic sensitivity analysis**

COV (%)				Limits (kPa)
$\gamma$	$\phi$	$\delta$	$T$	S
1	2	2	1	0 - 15
3	5	5	3	0 - 30
5	7	7	5	0 - 50
7	10	10	7	-
10	15	15	10	-

**Table 9. Sensitivity analysis of the COV of soil unit weight**

	COV (%)			Limits (kPa)	Results		
	$\gamma$	$\phi/\delta$	$T$	$S$	FS (mean)	PF (%)	$\beta$
Slope 5	1	7	5	0 - 30	1.489	0	4.153
	3	7	5	0 - 30	1.489	0	4.148
	5	7	5	0 - 30	1.489	0	4.141
	7	7	5	0 - 30	1.490	0	4.125
	10	7	5	0 - 30	1.491	0	4.094
Slope 6	1	7	5	0 - 30	1.519	0	3.847
	3	7	5	0 - 30	1.519	0	3.844
	5	7	5	0 - 30	1.519	0	3.838
	7	7	5	0 - 30	1.519	0	3.832
	10	7	5	0 - 30	1.520	0	3.817
Slope 8	1	7	5	0 - 30	1.487	0	3.853
	3	7	5	0 - 30	1.487	0	3.853
	5	7	5	0 - 30	1.487	0	3.852
	7	7	5	0 - 30	1.487	0	3.851
	10	7	5	0 - 30	1.488	0	3.845
Slope 9	1	7	5	0 - 30	1.497	0	4.640
	3	7	5	0 - 30	1.497	0	4.624
	5	7	5	0 - 30	1.497	0	4.587
	7	7	5	0 - 30	1.498	0	4.534
	10	7	5	0 - 30	1.500	0	4.426

Table 10. Sensitivity analysis of the COV of soil friction angle and soil-geogrid interface  
friction angle

	COV (%)			Limits (kPa)	Results		
	$\gamma$	$\phi/\delta$	$T$	$S$	FS (mean)	PF (%)	$\beta$
Slope 5	5	2	5	0 - 30	1.486	0	10.036
	5	5	5	0 - 30	1.488	0	5.537
	5	7	5	0 - 30	1.489	0	4.141
	5	10	5	0 - 30	1.493	0.0045	2.978
	5	15	5	0 - 30	1.501	1.2325	2.021
Slope 6	5	2	5	0 - 30	1.514	0	10.439
	5	5	5	0 - 30	1.516	0	5.228
	5	7	5	0 - 30	1.519	0	3.838
	5	10	5	0 - 30	1.525	0.0085	2.732
	5	15	5	0 - 30	1.539	1.6232	1.841
Slope 8	5	2	5	0 - 30	1.484	0	10.195
	5	5	5	0 - 30	1.485	0	5.220
	5	7	5	0 - 30	1.487	0	3.852
	5	10	5	0 - 30	1.491	0.0275	2.751
	5	15	5	0 - 30	1.500	2.0566	1.863
Slope 9	5	2	5	0 - 30	1.495	0	9.657
	5	5	5	0 - 30	1.496	0	5.983
	5	7	5	0 - 30	1.497	0	4.587
	5	10	5	0 - 30	1.500	0.0001	3.351
	5	15	5	0 - 30	1.507	0.4336	2.292

**Table 11. Sensitivity analysis of the COV of geogrid tensile strength**

	COV (%)			Limits (kPa)	Results		
	$\gamma$	$\phi/\delta$	$T$	S	FS (mean)	PF (%)	$\beta$
Slope 5	5	7	1	0 - 30	1.489	0	4.157
	5	7	3	0 - 30	1.489	0	4.150
	5	7	5	0 - 30	1.489	0	4.141
	5	7	7	0 - 30	1.489	0	4.122
	5	7	10	0 - 30	1.489	0	4.086
Slope 6	5	7	1	0 - 30	1.519	0	3.849
	5	7	3	0 - 30	1.519	0	3.846
	5	7	5	0 - 30	1.519	0	3.838
	5	7	7	0 - 30	1.519	0	3.830
	5	7	10	0 - 30	1.519	0	3.810
Slope 8	5	7	1	0 - 30	1.487	0	3.855
	5	7	3	0 - 30	1.487	0	3.852
	5	7	5	0 - 30	1.487	0	3.852
	5	7	7	0 - 30	1.487	0	3.849
	5	7	10	0 - 30	1.487	0	3.842
Slope 9	5	7	1	0 - 30	1.498	0	4.651
	5	7	3	0 - 30	1.498	0	4.629
	5	7	5	0 - 30	1.497	0	4.587
	5	7	7	0 - 30	1.497	0	4.527
	5	7	10	0 - 30	1.497	0	4.402

Table 12. Sensitivity analysis of the upper limit value of the surcharge load

	COV (%)			Limits (kPa)	Results		
	$\gamma$	$\phi/\delta$	$T$	S	FS (mean)	PF (%)	$\beta$
Slope 5	5	7	5	0 - 15	1.502	0	4.321
	5	7	5	0 - 30	1.489	0	4.141
	5	7	5	0 - 50	1.484	0	4.035
Slope 6	5	7	5	0 - 15	1.530	0	3.958
	5	7	5	0 - 30	1.519	0	3.838
	5	7	5	0 - 50	1.514	0	3.767
Slope 8	5	7	5	0 - 15	1.499	0	3.993
	5	7	5	0 - 30	1.487	0	3.852
	5	7	5	0 - 50	1.482	0	3.766
Slope 9	5	7	5	0 - 15	1.510	0	4.828
	5	7	5	0 - 30	1.497	0	4.587
	5	7	5	0 - 50	1.492	0	4.447

**Table 13. Sensitivity analysis for limit combinations**

	COV (%)			Limits (kPa)	Results		
	$\gamma$	$\phi/\delta$	$T$	S	FS (mean)	PF (%)	$\beta$
Slope 5	1	2	1	0 - 15	1.498	0	13.009
	10	15	10	0 - 50	1.497	1.3748	1.991
Slope 6	1	2	1	0 - 15	1.525	0	12.689
	10	15	10	0 - 50	1.536	1.7600	1.823
Slope 8	1	2	1	0 - 15	1.495	0	12.342
	10	15	10	0 - 50	1.496	2.1754	1.843
Slope 9	1	2	1	0 - 15	1.507	0	14.047
	10	15	10	0 - 50	1.504	0.6034	2.227

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# FIGURES

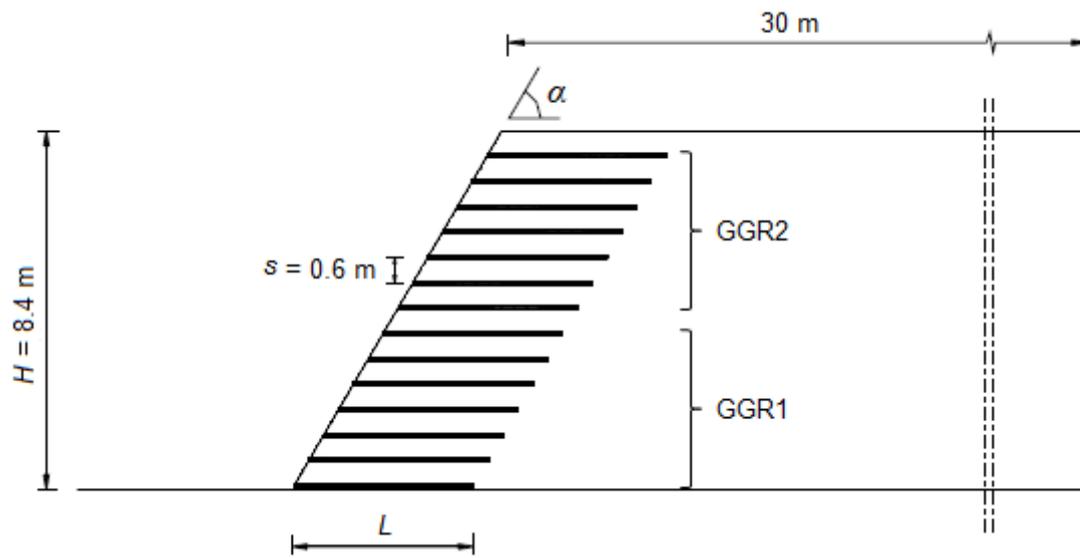


Figure 1. Schematic illustration of the geosynthetic-reinforced slopes

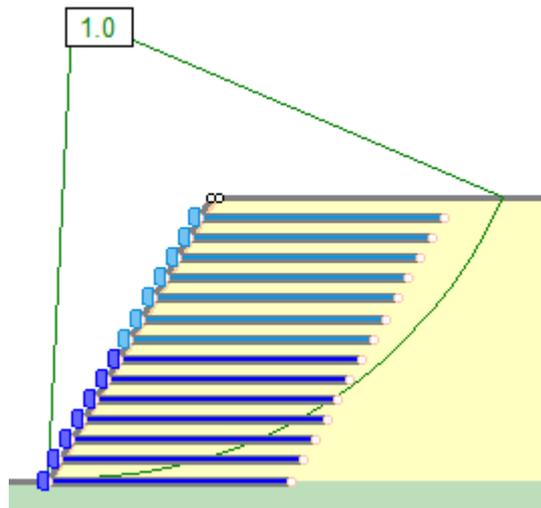


Figure 2. Global Minimum slip surface and over-design factor of Slope 1 ( $\alpha = 60^\circ$ ,  $L = 7.2$  m,  $\phi_d = 20^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

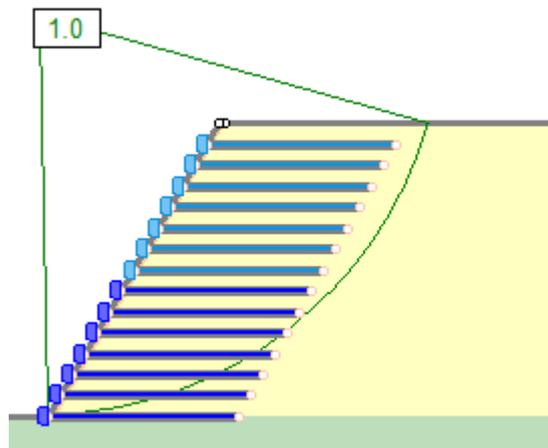


Figure 3. Global Minimum slip surface and over-design factor of Slope 2 ( $\alpha = 60^\circ$ ,  $L = 5.4$  m,

$$\phi_d = 25^\circ, \gamma_d = 23 \text{ kN/m}^3)$$

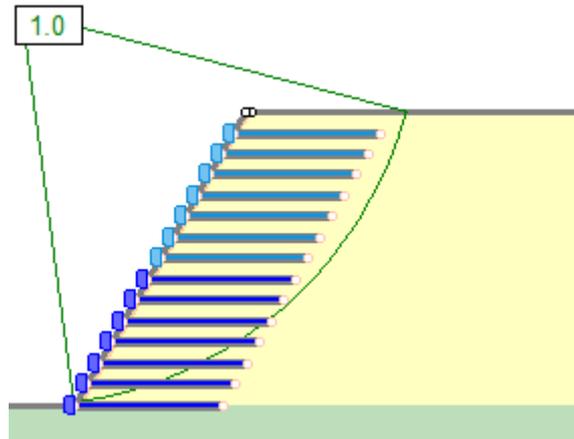


Figure 4. Global Minimum slip surface and over-design factor of Slope 3 ( $\alpha=60^\circ$ ,  $L=4.2$  m,

$$\phi_d=30^\circ, \gamma_d=23 \text{ kN/m}^3)$$

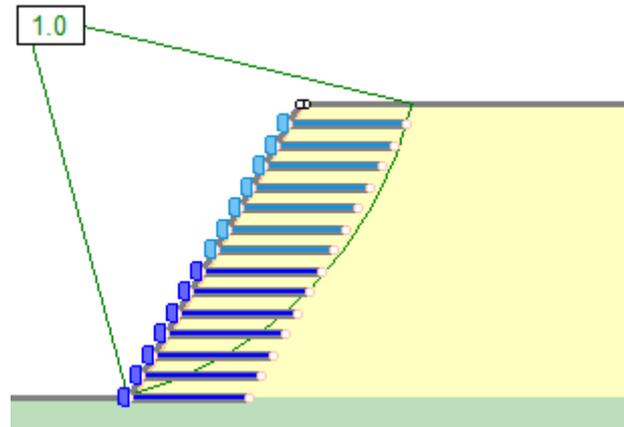


Figure 5. Global Minimum slip surface and over-design factor of Slope 4 ( $\alpha=60^\circ$ ,  $L=3.4$  m,

$$\phi_d=35^\circ, \gamma_d=23 \text{ kN/m}^3)$$

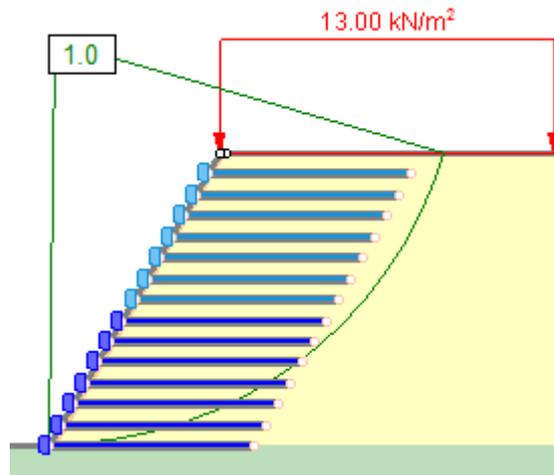


Figure 6. Global Minimum slip surface and over-design factor of Slope 5 ( $\alpha = 60^\circ$ ,  $L = 5.8 \text{ m}$ ,  
 $\phi_d = 25^\circ$ ,  $\gamma_d = 23 \text{ kN/m}^3$ )

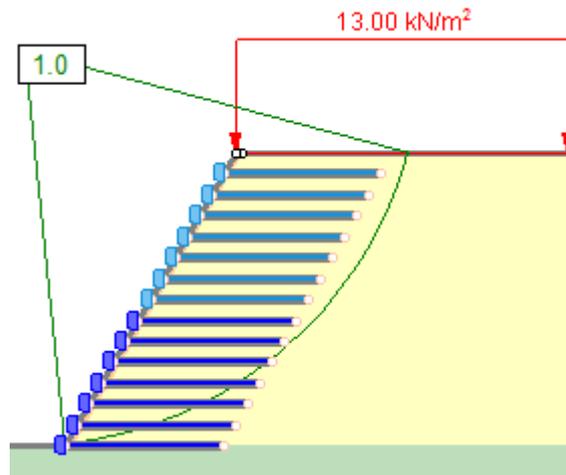


Figure 7. Global Minimum slip surface and over-design factor of Slope 6 ( $\alpha = 60^\circ$ ,  $L = 4.5$  m,  
 $\phi_d = 30^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

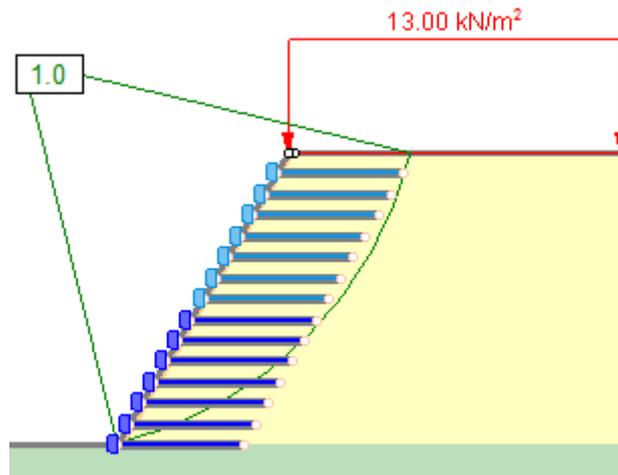


Figure 8. Global Minimum slip surface and over-design factor of Slope 7 ( $\alpha = 60^\circ$ ,  $L = 3.6$  m,

$$\phi_d = 35^\circ, \gamma_d = 23 \text{ kN/m}^3)$$

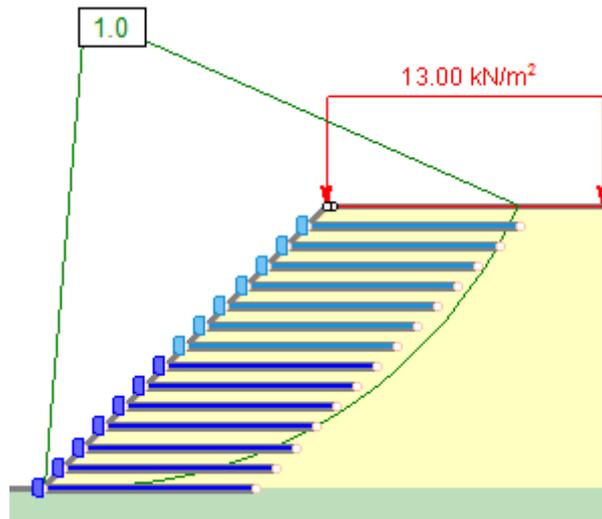


Figure 9. Global Minimum slip surface and over-design factor of Slope 8 ( $\alpha = 45^\circ$ ,  $L = 6.3$  m,  
 $\phi_d = 25^\circ$ ,  $\gamma_d = 23$  kN/m<sup>3</sup>)

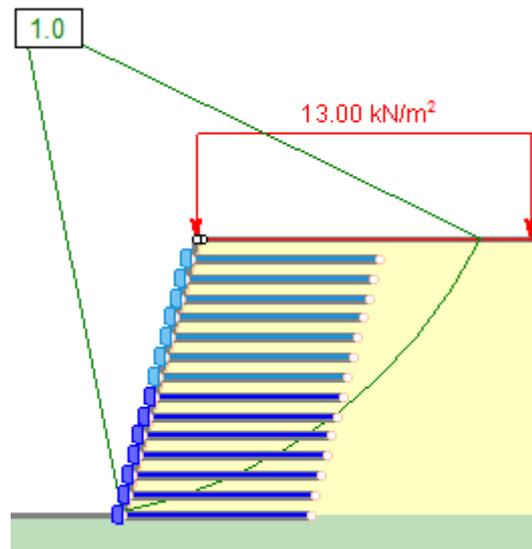


Figure 10. Global Minimum slip surface and over-design factor of Slope 9 ( $\alpha = 75^\circ$ ,

$$L = 5.8 \text{ m}, \phi_d = 25^\circ, \gamma_d = 23 \text{ kN/m}^3)$$

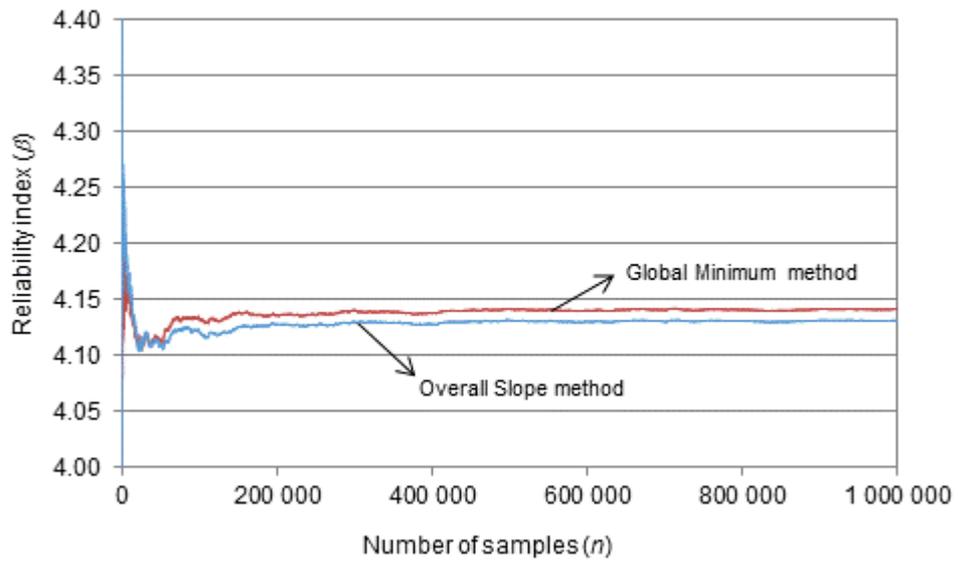
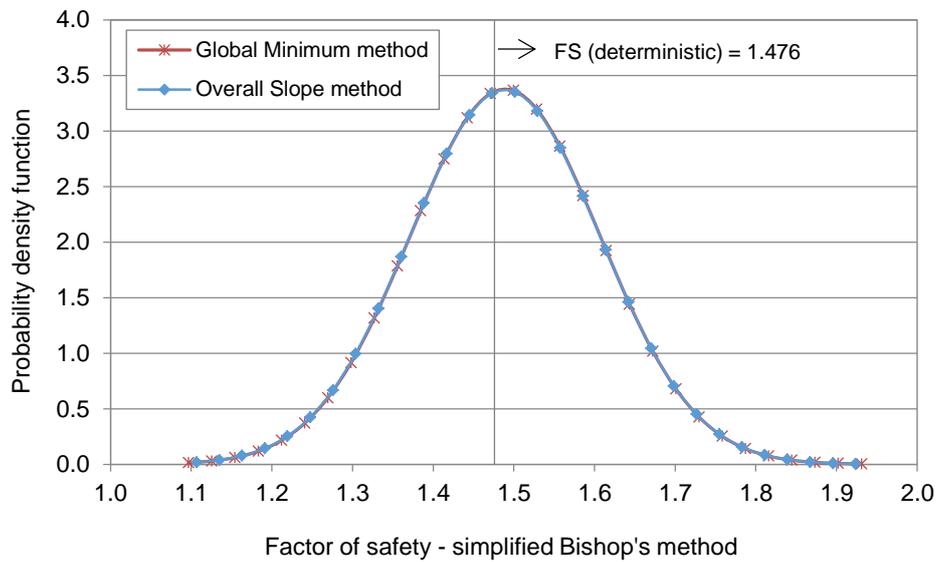
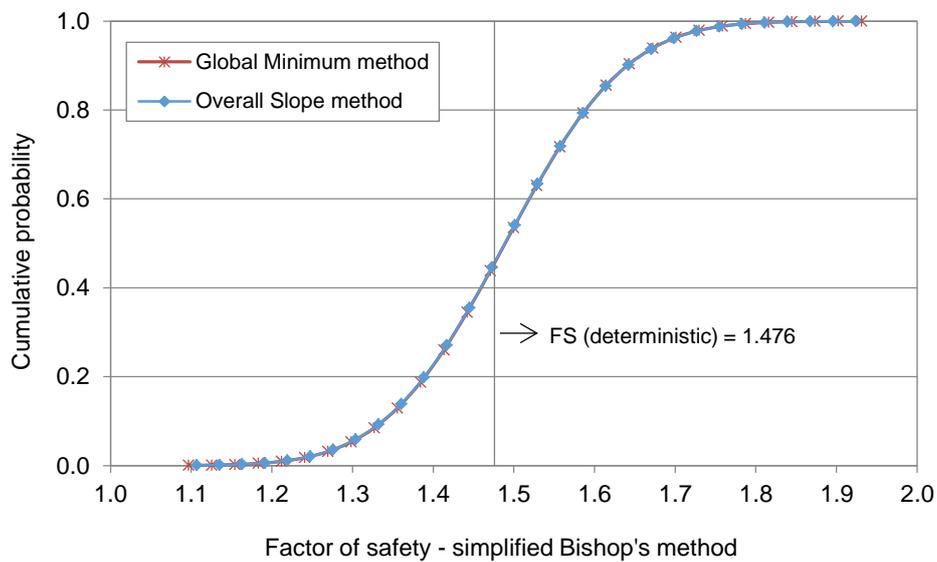


Figure 11. Convergence of the reliability index of Slope 5 (Global Minimum and Overall Slope methods)

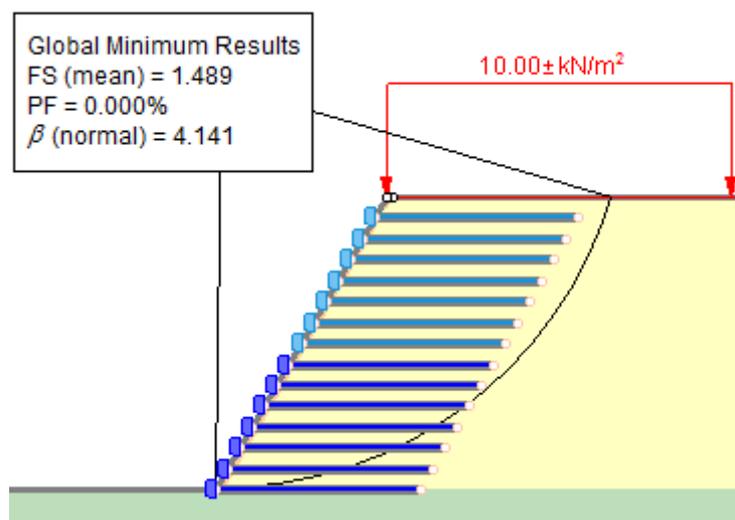


a)

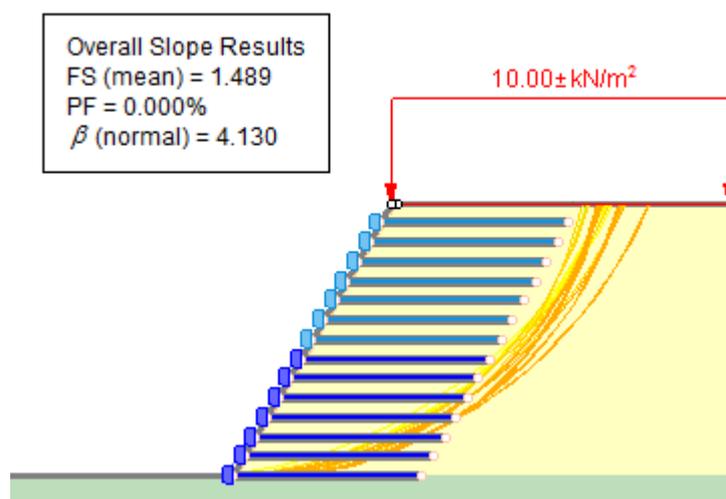


b)

Figure 12. Comparison of results of the probabilistic stability analysis of Slope 5 obtained from the Global Minimum and Overall Slope methods: a) probability density function of the factor of safety; b) cumulative probability distribution of the factor of safety



a)



b)

Figure 13. Results of the probabilistic stability analysis of Slope 5: a) Global Minimum method;

b) Overall Slope method