Modelling of Stability and Risk of Geotechnical Systems in Highly Variable Soils

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Synopsis: The paper will review the state-of-the-art in the use of finite element methods for modeling geotechnical engineering problems involving highly variable soil properties. Examples will focus on slope stability analyses in which traditional limit equilibrium methods, and even well-established risk assessment methodologies may lead to misleading results.

Keywords: Finite element method, Variable soils, Probability of failure, Random Fields, Risk assessment.

Introduction

The finite element method offers a powerful alternative to classical limit equilibrium methods of slope stability that have remained essentially unchanged for decades. The method offers the following main advantages:

- No assumption needs to be made in advance about the shape or location of the failure surface. The failure mechanism “seeks out” the weakest path through the soil.
- Since there is no concept of slices in the finite element approach there is no need for assumptions about slice side forces. The finite element method preserves global equilibrium until “failure” is reached.
- If realistic soil compressibility data is available, the finite element solutions will give information about deformations at working stress levels.
- The finite element method is able to monitor progressive failure up to and including overall shear failure.

Finite element slope stability analysis can hardly be considered a new technique. The first paper to tackle the subject by Smith & Hobbs (1) is over 35 years old followed by an important paper on the topic by Zienkiewicz et al. (2). Both of these papers had a very significant influence on the author’s finite element slope stability software developments over the years. Early publications date back to Griffiths (3) and the first ever published source code for finite element slope stability appeared in the second edition of the text by Smith & Griffiths in 1988 (4). Readers are also referred to Griffiths & Lane (5) for a thorough review of how the methodology works.
The paper will discuss risk assessment methods in geotechnical engineering, particularly for slope stability, including the most recent developments that combine random fields with finite element methods in the Random Finite Element Method (RFEM). Examples will be given of system slope reliability, where traditional methods may lead to quite misleading and unsafe results.

**Risk assessment in geotechnical engineering**

Soils and rocks are the most variable of all engineering materials, yet this is often coupled with inadequate site data. These factors combine to make geotechnical engineering one of the most appropriate areas for the application of probabilistic tools.

Risk assessment and probabilistic analyses in geotechnical engineering are rapidly growing areas of importance and activity for practitioners and academics e.g. Baecher and Christian (6), Fenton and Griffiths (7). At a recent G-I specialty conference called Georisk 2011 for example, several important state of practice papers were presented e.g. Christian and Baecher (8), Lacasse and Nadim (9), Scott (10). It is now commonplace for major geotechnical conferences to include sessions on risk assessment in geotechnical engineering.

Of all areas of geotechnical engineering, slope stability analysis has received greater attention using risk assessment tools than any other, since the concept of replacing a “factor of safety” by a “probability of failure” is immediately appealing to many engineers, see e.g. Alonso (11), Catalan and Cornell (12), Li and Lumb (13), Oka and Wu (14), Chowdhury and Xu (15), Mostyn and Soo (16), Juang et al. (17), Mostyn and Li (18), Lacasse (19), Lacasse and Nadim (20), Liang et al (21), Malkawi et al. (22), Griffiths and Fenton (23,24), Duncan (25), El Ramly et al. (26), Bhattacharya et al. (27), Babu and Mukesh (28), Jimenez-Rodriguez et al. (29), Low and Tang (30), Hong and Roh (31), Griffiths et al. (32), Huang et al. (33), Ching et al. (34), Mbarka et al. (35), Wang et al. (36).

**The Random Finite Element Method (RFEM)**

The goal of a probabilistic slope stability analysis is to estimate the probability of slope failure as opposed to the ubiquitous factor of safety used in conventional analysis. Several relatively simple tools exist for performing this calculation that include the First Order Second Moment (FOSM) method and the First Order Reliability Method (FORM). The FORM method in particular has now been developed to a quite significant level of sophistication to tackle correlation and system slope reliability, e.g. Low et al. (37), Low et al. (38).

A legitimate criticism of these first order methods however, is that they are unable to properly account for spatial correlation in the 2D or 3D random materials, and are inextricably linking with “old fashioned” slope stability methods that involve simple shapes for the failure surfaces.

To overcome these deficiencies, a method called the Random Finite Element Method (RFEM) that combines random field theory with deterministic finite element analysis was developed by the authors in the early 1990’s and has been applied to a wide range of geotechnical applications, e.g. Griffiths and Fenton (39), Fenton and Griffiths (7). In a stability analysis, input to RFEM is provided in the form of the mean, standard deviation and spatial correlation length of the soil strength parameters at the “point” level, which may consist of several layers with different statistical input parameters. In the absence of site specific information, there is an increasing number of publications presenting typical ranges for the
standard deviation of familiar soil properties, e.g. Lee et al. (40).

In RFEM, local averaging is fully accounted for at the element level indicating that the mean and standard deviation of the soil properties are statistically consistent with the mesh density. Since the finite element method of slope stability allows mechanisms to “seek out” the most critical path through the soil, the method offers great promise for more realistic reliability assessment of slopes and other geotechnical applications. The flow chart for a typical RFEM slope stability analysis is shown in Figure 1.

![Flow chart for a typical RFEM slope stability analysis](image)

Fig. 1. Flow chart for a typical RFEM slope stability analysis.

The RFEM codes developed by Griffiths and Fenton for a range of geotechnical applications are freely available in source code from the authors’ web site at www.mines.edu/~vgriffit/rfem. The 2D slope stability program is called rslope2d. A couple of failure mechanisms computed using this program for slopes with quite different spatial correlation lengths but with the same mean and standard deviation of strength parameters are shown in Figure 2. The spatial correlation length (assumed isotropic) is expressed in dimensionless form relative to the height of the embankment, e.g. $\Theta_c = 0.5$ means the spatial correlation length is $0.5H$ etc. It is seen that the slope with the higher spatial correlation length in the lower figure gives a quite smooth failure mechanism, more like the classical “mid-point” circle. The soil with a lower spatial correlation length in the upper figure however, displays a quite complex system of interacting mechanisms which would defy analysis by any traditional LEM.
Following the results of Griffiths and Fenton (24), the RFEM results for an undrained clay slope with a spatially random, lognormally distributed dimensionless undrained strength given by \( C = c_u / (\gamma_{sat} H) \) is shown in Figure 3. The computed probability of failure by RFEM \( (p_f) \) is given as a function of the spatial correlation length \( (\Theta_C = \theta_{inc}/H) \) and the coefficient of variation \( (V_C = \sigma_C / \mu_C) \). It can be seen that an increasing correlation length may either increase or decrease the slope failure probability depending on the input coefficient of variation \( V_C \).

![Figure 2](image1.jpg)

**Fig. 2.** Typical failure mechanisms from an RFEM analysis with two different spatial correlation lengths.

![Figure 3](image2.jpg)

**Fig. 3.** Influence of the spatial correlation length and coefficient of variation on the probability of failure of an undrained slope \((\mu_C = 0.25)\).

In order to interpret these results, a couple of key deterministic solutions considering a homogeneous soil should be kept in mind. (i) if \( \mu_C = 0.25 \), \( FS = 1.47 \) and (ii) if \( \mu_C = 0.17 \), \( FS = 1.0 \). The diverging
results in Figure 3 can then be explained by considering the limiting cases of $\Theta_c \to 0$ and $\Theta_c \to \infty$. As $\Theta_c \to 0$, the local averaging is maximized, and the slope becomes essentially homogeneous with a uniform strength given by the median of the strength distribution. If the median falls below 0.17, all simulations fail and $p_f \to 1$. On the other hand, if the median is greater than 0.17, none of the simulations fail and $p_f \to 0$. As $\Theta_c \to \infty$, each simulation involves a uniform soil with the property varying from one simulation to the next, so $p_f \to P[C < 0.17]$.

For example, in the case of $\mu_c = 0.25$, $V_c = 0.5$, the parameters of the underlying normal distribution of $\ln C$ are given as

$$\mu_{\ln C} = \ln \mu_x - \frac{1}{2} \ln \{1 + V_c^2\} = -1.498$$

$$\sigma_{\ln C} = \sqrt{\ln \{1 + V_c^2\}} = 0.472$$

hence

$$p_f = \Phi\left(\frac{\ln 0.17 - \mu_{\ln C}}{\sigma_{\ln C}}\right) = 0.28$$

which is shown as the asymptotic trend of the line corresponding to $V_c = 0.5$ as $\Theta_c \to \infty$ in Figure 3.

As $\Theta_c \to 0$ however, the median of the shear strength is given by

$$\text{Median}_c = \exp(\mu_{\ln C}) = \exp(-1.498) = 0.22 > 0.17$$

hence $p_f \to 0$.

First order methods and single random variable Monte-Carlo methodologies that treat each simulation as a homogeneous material, can be considered special cases of RFEM with $\Theta_c \to \infty$ but cannot be guaranteed to deliver conservative results.

**Influence of Mesh Refinement**

A commonly asked question of any finite element analysis, including RFEM, is the extent to which mesh refinement and discretization errors affect the results. As mentioned previously, the statistics of the random field mapped onto the finite element mesh are adjusted in a consistent way to account for element size. This is an integral part of the Local Average Subdivision method, Fenton and Vanmarcke (41). As for the overall discretization issue, Figure 4 shows the influence of mesh refinement for two different cases. It can be seen that the finer mesh gives somewhat higher values of $p_f$, which is to be expected, since more paths are available for failure to occur.
Fig. 4. Influence of mesh density on $p_f$ for an undrained slope with $\Theta_c = 1$ and $\mu_c = 0.25$.

**Importance of spatial variability**

The infinite slope problem shown in Figure 5 is one of the oldest and simplest types of slope problem in which the failure mechanism is assumed to be purely translational with the failure plane at the base of the layer. In the absence of pore pressures ($u = 0$), the factor of safety can be expressed explicitly by the equation

$$\tan \sin \cos \tan c_{FS} H \phi \beta \beta = +$$

(4)

In this example as discussed by Griffiths et al. (42), the cohesion is defined by $\mu_c = 10\, \text{kN/m}^2$ and $\sigma_c = 3.0\, \text{kN/m}^2$ and the tangent of the friction angle by $\mu_{\tan \phi'} = 0.5774$ and $\sigma_{\tan \phi'} = 0.1732$. The remaining parameters are assumed to be deterministic with values given by $H = 5.0\, \text{m}$, $\beta = 30^\circ$, and $\gamma = 17.0\, \text{kN/m}^3$. Substitution of these deterministic parameters and the mean values of the random variables into Eq. (4) leads to a deterministic factor of safety of $FS = 1.27$.

From Eq. (4), and assuming $c'$ and $\tan \phi'$ are uncorrelated, we can estimate the mean and standard deviation of $FS$ by the FOSM as
\[ \mu_{FS} = \frac{\mu_c}{\gamma H \sin \beta \cos \beta} + \frac{\mu_{\tan \phi'}}{\tan \beta} \]

\[ \sigma_{FS} = \sqrt{\left(\frac{1}{\gamma H \sin \beta \cos \beta}\right)^2 \sigma_c^2 + \left(\frac{1}{\tan \beta}\right)^2 \sigma_{\tan \phi'}^2} \]

which gives \( \mu_{FS} = 1.27 \) and \( \sigma_{FS} = 0.311 \)

Assuming that \( FS \) is lognormal, the probability of failure is then given by

\[ p_f = P[FS < 1] = P[\ln(FS) < \ln(1)] = \Phi\left[-\frac{\mu_{\ln FS}}{\sigma_{\ln FS}}\right] \]

where the mean and standard deviation of the underlying normal distribution of \( \ln(FS) \) are given from Eq.(1) as \( \mu_{\ln(FS)} = 0.2113 \) and \( \sigma_{\ln(FS)} = 0.2409 \). After substitution

\[ p_f = \Phi\left[-0.2113 \right] = 1 - \Phi\left[0.2409\right] = 1 - 0.810 = 0.19 \]

hence the probability of failure is approximately 19\%. It should be noted that this result, being based on the deterministic Eq.(4), assumes failure always occurs at the base of the layer.

The same problem was then solved using RFEM by including lognormal and uncorrelated \( c' \) and \( \tan \phi' \) and a range of spatial correlation lengths defined in dimensionless form as \( \Theta = \theta/H \) (assumed in this example to be the same for both \( c' \) and \( \tan \phi' \)). The results shown in Figure 6 indicate that the FORM results are consistently unconservative, but less so as \( \Theta \to \infty \).

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Fig. 6. Comparison of RFEM and FORM results for an infinite slope analysis.
This is because in RFEM, failure takes place along the weakest path, which doesn’t necessarily occur at the base of the layer. For shorter values of $\Theta$, the critical plane is more likely to occur above the base and $p_f$ is higher. The figure also shows a typical random field and failure plane from the RFEM Monte-Carlo analyses.

**System reliability of slopes**

**Deterministic analysis**

A $26.6^\circ (2:1)$ undrained ($\phi_u = 0$) slope is considered with the slope profile shown in Fig. 7a. The slope has height $H = 10.0 \text{ m}$, soil unit weight $\gamma_{sat} = 20.0 \text{ kN/m}^3$, shear strength $c_{u1} = 30.6 \text{ kPa}$ (expressed in a dimensionless form given by $C_{u1} = c_{u1} / (\gamma_{sat}H) = 0.153$). The $FS$ of the slope was found to be 1.25 and the deformed mesh at failure is shown in Fig. 7b.

Another two-layer slope with a similar geometry, but including a foundation with depth ratio $D = 2$ is shown in Fig. 8a. The foundation was assumed to be undrained soil with the same unit weight, but with a different shear strength given by $c_{u2} = 45.8 \text{ kPa}$ ($C_{u2} = 0.229$). The $FS$ of the two-layer slope was found to be also 1.25 with the deformed mesh at failure shown in Fig. 8b. As shown by Griffiths and Lane (5) for this case, if $C_{u2} / C_{u1} \geq 1.5$, the foundation strength has no influence on the $FS$ as shown in Figure 9. A deep mechanism is observed when $C_{u2} / C_{u1} \leq 1.5$, whereas a shallow mechanism is seen when $C_{u2} / C_{u1} \geq 1.5$. At the transition or bifurcation point when $C_{u2} / C_{u1} = 1.5$, both mechanisms are trying to form at the same time as shown in Fig. 8b.
Probabilistic analysis

For an undrained slope without a foundation as considered by Huang et al. (33), if the shear strength is treated as a single random variable ignoring spatial variability (FORM), \( p_f \) is simply equal to the probability that the shear strength parameter \( C_{u1} \) will be less than \( C_{u1,FS=1} \), where \( C_{u1,FS=1} \) is the value that results in \( FS = 1 \). Quantitatively, this equals the area beneath the probability density function corresponding to \( C_{u1} \leq C_{u1,FS=1} \). For the slope shown in Fig. 7a, \( C_{u1,FS=1} = 0.122 \) and \( C_{u1,FS=1.25} = 0.153 \), so if we let \( \mu_{C_{u1}} = 0.153 \) and \( \sigma_{C_{u1}} = 0.046 \) (\( v_{C_{u1}} = 0.3 \)), Eqs. (1) give that the mean and standard deviation of the underlying normal distribution are \( \mu_{\ln C_{u1}} = -1.920 \) and \( \sigma_{\ln C_{u1}} = 0.294 \), hence

\[
p_f = p[C_{u1} < 0.122] = \Phi \left( \frac{\ln 0.122 - \mu_{\ln C_{u1}}}{\sigma_{\ln C_{u1}}} \right) = 0.266 \tag{9}
\]

For the undrained two-layer slope shown in Fig. 8a, the FORM method combined with response surface method ignoring spatial variability was used to calculate \( p_f \). By changing \( \mu_{C_{u2}} / \mu_{C_{u1}} \) in the range of \( \{0.25, 0.5, \ldots, 2.5\} \) and fixing \( \mu_{C_{u1}} = 0.153 \) and \( v_{C_{u2}} = v_{C_{u1}} = 0.3 \), the influence of the strength of the foundation on the \( p_f \) was investigated with results shown in Fig. 10.

Also shown in Fig. 10 are the same analyses performed by RFEM including spatial variability \( \Theta_{\ln C_{u2}} = \Theta_{\ln C_{u1}} = 0.5 \), and the “embankment only” result \( p_f = 0.071 \) which is for the slope shown in Fig. 7a treating \( C_{u1} \) as a random variable with statistical strength parameters \( \mu_{C_{u1}} = 0.153 \), \( v_{C_{u1}} = 0.3 \) and \( \Theta_{\ln C_{u1}} = 0.5 \).

The foundation strength had little influence on \( p_f \) for the two-layer slope if \( \mu_{C_{u2}} / \mu_{C_{u1}} > 1.50 \) by both RFEM and FORM (ignoring spatial variability). When \( \mu_{C_{u2}} / \mu_{C_{u1}} = 1.50 \), RFEM gave a higher \( p_f = 0.118 \) for the two-layer slope than the \( p_f = 0.071 \) in the “embankment only” case which has only one mechanism as shown in Fig. 7b. In other words, RFEM accurately predicts the system probability of failure, but FORM (ignoring spatial variability) only catches the failure mechanism with the highest \( p_f = 0.226 \). Although FORM is more conservative than RFEM in this example, Griffiths et al. (32) and Huang et al. (33) discuss other combinations where the opposite is true.
Three-dimensional slope reliability

Since the 2-d factor of safety is generally considered to be conservative, practitioners are reluctant to invest in the more time-consuming 3-d approaches. A key question to be addressed is, under what circumstances will the probability of failure of a slope predicted by a full 3-d analysis be higher than that obtained from an equivalent 2-d analysis?

In all the RFEM analyses that follow from Griffiths et al. (43) and referring to Fig. 11, the bottom of the mesh (y = H) is fully fixed and the back of the mesh (x = 0) is allowed to move only in a vertical plane. Both “rough” and “smooth” boundary conditions have been considered at the ends in the out-of-plane direction (z = 0 and L). In the rough cases the ends are fully fixed and in the smooth case, they are allowed to move only in a vertical plane. In this study, it was determined that 2000 realizations of the Monte-Carlo process for each parametric group, was sufficient to give reliable and reproducible estimates of the probability of failure $p_f$.

The undrained clay slope at failure from a typical simulation shown in Figure 11 demonstrates an important characteristic in 3D slope analysis called the “preferred” failure mechanism width $W$. This is the width of the failure mechanism in the $z$-direction that the finite element analysis “seeks out”. Over a suite of Monte-Carlo simulations the average preferred failure mechanism width is called $W_{\text{crit}}$. It will be shown that this dimension has a significant influence on 3D slope reliability depending on whether the length of the slope $L$ is greater than or less than $W_{\text{crit}}$.

The length ratio has been varied in the range $0.2 < L/H < 16$ to investigate the influence of three-dimensionality, with results presented in Figure 12. In the case of smooth boundary conditions, the $p_f$ of one slice ($L/H = 0.2$) in the 3-d analysis is equivalent to that given by a 2D RFEM analysis since the 3D analysis is essentially replicating plane strain. It is also shown in the smooth case that as $L/H$ is increased, $p_f$ initially decreases, reaching a minimum before rising to eventually exceed the 2D value. In the rough case, $p_f$ is close to zero for a narrow slice and increases steadily as $L/H$ is increased due to a gradual reduction in the supporting influence of the rough boundaries in the 3D case.
As the length ratio is increased in both the rough and smooth cases, the 3-d $p_f$ eventually exceeds the 2D value, indicating that 2D analysis will be always give unconservative results if the slope is long enough.

It may also be speculated that $p_f \to 1$ as $L/H \to \infty$ regardless of boundary conditions.

Fig. 12. Probability of failure versus slope length ratio

$(V_{cu} = 0.5, \theta = H, \ FS = 1.39, \ slope\ angle\ 2h:1v)$
For the case of smooth boundary conditions, let us define the critical slope length \( L_{\text{crit}} \) and the critical slope length ratio \( \left( \frac{L}{H} \right)_{\text{crit}} \) as being that value of \( L/H \) for which the slope is safest and its probability of failure \( p_f \) a minimum. It will be shown that this minimum probability of failure in the smooth case occurs when \( L_{\text{crit}} \approx W_{\text{crit}} \). If we reduce the slope length ratio below this critical value \( L < L_{\text{crit}} \), the slope finds it easier to form a global mechanism spanning the entire width of the mesh with smooth end conditions, so the value of \( p_f \) increases, tending eventually to the plane strain value. However, if we increase the slope length ratio above this critical value \( L > L_{\text{crit}} \), the slope finds it easier to form a local mechanism. Since \( L > W_{\text{crit}} \) the mechanism has more opportunities to develop somewhere in the \( z \)-direction hence \( p_f \) again increases.

**Concluding remarks**

The paper has demonstrated the power and advantages of the finite element method for probabilistic slope stability analysis in highly variable soils. Examples of slope risk analysis were presented using the random finite element method (RFEM) developed by the authors. It was shown that single random variable approaches can give unconservative results compared with RFEM using 2D random fields. The key benefit of RFEM is that it does not require any \textit{a priori} assumptions related to the shape or location of the failure mechanism. In an RFEM analysis, the failure mechanism has freedom to “seek out” the weakest path through the random soil, which generally leads to more simulations reaching failure. The importance of spatial variability was further demonstrated in an example of system slope stability risk analysis, and two examples involving an infinite slope and a 3D slope. In both cases, failure to account for spatial variability could lead to unconservative results.

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**References**


