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Probabilistic Analysis of Shallow Foundations on Rocks Obeying Hoek-Brown Failure Criterion

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ABSTRACT

A probabilistic analysis using Polynomial Chaos Expansion method is presented to compute the probability density function (PDF) of the ultimate bearing capacity of a strip footing resting on a rock mass and subjected to a vertical/inclined load. The rock is assumed to follow Hoek-Brown failure criterion. The kinematic approach of the limit analysis theory is used. The results of the vertical load case have shown that: (i) the Geological Strength Index GSI and the uniaxial compressive strength of the intact rock $\sigma_c$ have the most significant weight in the variability of the ultimate bearing capacity, (ii) the non-normality of the input variables has a significant effect on the shape of the PDF of the ultimate bearing capacity and, (iii) a negative correlation between GSI and $\sigma_c$ leads to less spread out PDF. Finally, it was shown in the inclined load case that the variability of the ultimate bearing capacity decreases with the increase of the footing load inclination.

INTRODUCTION

Traditionally, the stability analysis of shallow foundations resting on rocks is based on deterministic approaches. The present paper aims at determining the ultimate bearing capacity of a centrally loaded shallow strip footing (i.e. a footing subjected to a vertical or an inclined load) using a probabilistic analysis. The footing rests on a rock mass that follows the modified Hoek-Brown failure criterion. The deterministic models are based on the kinematic approach of limit analysis theory using translational multiblock failure mechanisms. The Polynomial Chaos Expansion (PCE) method (Huang et al. 2009) is used for the probabilistic analysis. Four uncertain parameters related to the modified Hoek-Brown failure criterion are modeled as random variables. These are the Geological Strength Index (GSI), the uniaxial compressive strength of the intact rock ($\sigma_c$), the intact rock material constant ($m_i$) and the disturbance coefficient ($D$). The paper is organized as follows: The first section aims at presenting the basic idea of the PCE methodology. It is followed by a presentation of the deterministic model used for the computation of the ultimate
bearing capacity. Finally, the probabilistic results are presented and discussed. The paper ends with a conclusion.

POLYNOMIAL CHAOS EXPANSION (PCE) METHODOLOGY

In the PCE methodology, the response of the mechanical model $\Gamma$ with $M$ input random variables is expressed by a PCE of order $p$ pre-defined by the user as follows:

$$\Gamma_{PCE}(\xi) = \sum_{\beta=0}^{\infty} a_{\beta} \Psi_{\beta}(\xi) \approx \sum_{\beta=0}^{p-1} a_{\beta} \Psi_{\beta}(\xi)$$

(1)

where $\xi$ are independent standard normal random variables representing the input random variables, $a_{\beta}$ are unknown coefficients to be determined, $\Psi_{\beta}$ are chosen as multidimensional Hermite polynomials and $P = (M + p)!/(M! p!)$ is the number of terms retained in the truncation scheme (which is equal to the number of unknown coefficients). To compute the coefficients $a_{\beta}$ in Equation 1, two non intrusive methods have been proposed in literature: the projection and the regression approach. In this paper, the regression approach (Huang et al. 2009) is used. In this approach, each input random variable is assumed to take the roots of the one-dimensional Hermite polynomial (of one degree higher than the PCE order $p$). For $M$ input random variables and a PCE of order $p$, the available collocation points (sample points) are the result of all possible combinations of these roots. Their number $N = (p + 1)^M$ dramatically increases as $p$ and $M$ increase and it is always higher than the number $P$ of the unknown coefficients (for $M \geq 2$). This leads to a linear system of equations whose number of equations is greater than the number of unknowns. Several attempts have been made in literature to select the most efficient collocation points among the $N$ available ones in order to reduce the number of calls to the deterministic model. The approach proposed by Sudret (2008) is used herein. In order to check the goodness of fit of the PCE, the coefficient of determination $R^2$ is computed. The value $R^2 = 1$ indicates a perfect fit of the true model response $\Gamma$, whereas $R^2 = 0$ indicates a nonlinear relationship between the true model $\Gamma$ and the PCE model $\Gamma_{PCE}$. Once the output approximation via a PCE is obtained, this PCE will be called meta-model and will be employed for the probabilistic analyses. The PDF and the statistical moments of the system response can be estimated. This can be done by simulating a large number of realizations of the standard normal variables on the meta-model using Monte Carlo technique. Another important advantage of the meta-model is that a global sensitivity analysis (GSA) based on Sobol indices can be performed analytically using the coefficients of the PCE (Sudret 2008). These indices indicate the contribution of each random variable or combination of random variables to the variability of the system response. This is important because it helps engineers in detecting the input uncertain parameters which have a significant influence in the variability of the system response.
DETERMINISTIC BEARING CAPACITY OF A CENTRALLY LOADED STRIP FOOTING

The modified Hoek-Brown failure criterion is used in this paper. This failure criterion only deals with intact rocks or heavily jointed rock masses. A heavily jointed rock mass involves sufficiently dense and randomly distributed joints so that in the scale of the problem, it can be regarded as an isotropic assembly of interlocking particles. Consequently, rocks with few discontinuities cannot be considered in this framework. This failure criterion can be described by the following equation (Hoek et al. 2002):

$$\sigma_1 - \sigma_3 = \sigma_c \left[m \left(\frac{\sigma_3}{\sigma_c}\right) + s\right]^n$$  \hspace{1cm} (2)

where $\sigma_1$ and $\sigma_3$ are respectively the major and minor principal stresses at failure and $\sigma_c$ is the uniaxial compressive stress of the rock at failure. The parameters $m$, $s$ and $n$ are given by the following equations:

$$ m = m_i e^{\left(\frac{GSI-100}{28-14D}\right)}; \quad s = e^{\left(\frac{GSI-100}{9-3D}\right)}; \quad n = (1/2) + (1/6) \left[e^{-\frac{GSI}{15}} - e^{-20/3}\right]$$  \hspace{1cm} (3)

In these equations, the geological strength index ($GSI$) characterizes the quality of rock masses, and it depends on the rock mass structure as well as the surface condition of joints (Hoek et al. 2002). On the other hand, the parameter $m_i$ is the value of parameter $m$ for intact rock and can be obtained from experimental tests. It varies from 4 for very fine weak rock like claystone to 33 for coarse igneous light-colored rock like granite. Finally, $D$ is the disturbance coefficient. It varies from 0.0 for undisturbed in situ rock masses to 1.0 for very disturbed rock masses.

Two kinematically admissible failure mechanisms M1 and M2 based on the upper-bound theorem of limit analysis are used herein. They are described in Soubra (1999) in the case of a Mohr-Coulomb material whose failure criterion is linear. M1 is a translational symmetrical multiblock failure mechanism (Figure 1a) and is used for the computation of the ultimate bearing capacity of a vertically loaded strip footing. It is composed of $2k+1$ triangular rigid blocks (a central symmetrical block under the footing and $2k$ symmetrical rigid blocks at both sides of the footing). This mechanism can be completely described by $2k$ angular parameters which are $\alpha_i (i=1, \ldots, k-1)$, $\beta_i (i=1, \ldots, k)$ and $\theta$. On the other hand, M2 is a translational non-symmetrical multiblock failure mechanism (Figure 1b) and is suitable for the calculation of the ultimate bearing capacity of obliquely loaded strip foundations. This mechanism is composed of $k$ triangular rigid blocks. It can be completely described by $2k-1$ angular parameters which are $\alpha_i (i=1, \ldots, k-1)$ and $\beta_i (i=1, \ldots, k)$.

For the case of rock masses obeying the modified Hoek-Brown non linear failure criterion, Yang and Yin (2005) have replaced this criterion by an equivalent linear Mohr-Coulomb failure criterion (Figure 2). Recently, a more rigorous and efficient approach which preserves the original non linear form of the modified Hoek-Brown failure criterion, was proposed by Saada et al. (2008). A similar approach is used herein for both M1 and M2 mechanisms. Notice that the energy dissipation takes
place at the discontinuity surfaces $d_i$ between the material at rest and the material in motion, and at the discontinuity surfaces $l_i$ within the radial shear zone (Figure 1). In this paper, the inter-block velocities $v_{i,i+1}$ will be assumed as being inclined at angles $\phi_{i,i+1}$ to lines $l_{i+1}$ (Figure 3) where $\phi_{i,i+1}$ will be different along the different lines $l_{i+1}$. Also, the block velocities $v_{i+1}$ will be assumed as being inclined at angles $\phi_{i+1}$ to lines $d_{i+1}$ (Figure 3) where $\phi_{i+1}$ will be also different along the different lines $d_{i+1}$.

Figure 1. Failure mechanisms for the computation of the bearing capacity in the case of a) vertical load and b) inclined load.

Figure 2. Failure envelope the Hoek-Brown and the equivalent mechanism Mohr-Coulomb failure criteria.

Figure 3. Velocity field of M1 mechanism.
For both M1 and M2 mechanisms, it was found that an upper-bound of the ultimate bearing capacity can be defined as:

\[
q_u = \frac{1}{2} \gamma B_0^2 N'_y + qB_0 N'_q + \sigma_c B_0 N'_{\sigma_c} 
\]  

(4)

where \(N'_y, N'_q\) and \(N'_{\sigma_c}\) are non-dimensional functions which can be expressed in terms of the mechanism geometrical parameters and in terms of the tangential frictional angle \(\varphi_{i+1}\) and \(\varphi_{i,i+1}\) where \(i=0, \ldots, k-1\). For each failure mechanism, the ultimate bearing capacity is obtained by minimization with respect to the angular parameters of the failure mechanism and with respect to \(\varphi_{i+1}\) and \(\varphi_{i,i+1}\).

**PROBABILISTIC NUMERICAL RESULTS**

The aim of this section is to present the probabilistic results of the ultimate bearing capacity. The footing of width \(B=1\text{m}\) is placed on a weightless \((\gamma=0)\) rock mass, with no surcharge loading on the ground surface \((q=0)\). The illustrative values of the statistical moments of the random variables as well as the correlation coefficient are presented in Table 1 [Hoek 1998 and Hoek et al. 2002].

<table>
<thead>
<tr>
<th>Variables</th>
<th>Mean</th>
<th>COV</th>
<th>Probability distribution</th>
<th>Coefficient of correlation (\rho)</th>
</tr>
</thead>
<tbody>
<tr>
<td>GSI [-]</td>
<td>25</td>
<td>10%</td>
<td>Log-normal</td>
<td>(-0.75 \leq \rho(GSI, \sigma_c) \leq +0.75)</td>
</tr>
<tr>
<td>m_i [-]</td>
<td>8</td>
<td>12.5%</td>
<td>Log-normal</td>
<td></td>
</tr>
<tr>
<td>(\sigma_c) [MPa]</td>
<td>10</td>
<td>25%</td>
<td>Log-normal</td>
<td></td>
</tr>
<tr>
<td>D [-]</td>
<td>0.3</td>
<td>10%</td>
<td>Log-normal</td>
<td></td>
</tr>
</tbody>
</table>

**CASE OF A VERTICAL LOAD**

*Optimal PCE order and Sobol indices*

The coefficient of determination \(R^2\) of the PCEs of orders 2, 3, 4 and 5 were computed. It was found that \(R^2\) of order 4 and order 5 are close to 1 (i.e. higher than 0.999999). Consequently a PCE of order \(p=4\) provides a good fit between the meta-model and the true model, and will be adopted in the rest of the probabilistic studies performed in this paper. For the prescribed values of the statistical moments of the input random variables presented in Table 1, the Sobol indices have been computed. The Sobol indices of parameter \(\sigma_c\) (0.5378) and \(GSI\) (0.3141) are much higher than those of parameters \(m_i\) (0.0866) and \(D\) (0.0308). A thorough experimental investigation will thus be required by the engineer for only the influential parameters (i.e. \(GSI\) and \(\sigma_c\)) to obtain reliable results of the system response.

*Effect of the coefficient of variation (COV) of the random variables*

The effect of COV of the random variables is presented in Table 2. This table shows that the COV has practically no effect on the mean value of the ultimate bearing capacity and that the increase in the COV of parameters \(GSI\) or \(\sigma_c\) has a significant effect on the variability of the ultimate bearing capacity. Table 2 also
shows the effect of COV on Sobol indices. One can note that the increase in COV of a certain variable increases its Sobol index and decreases the Sobol indices of the other variables.

Table 2. Effect of the coefficients of variation of the input random variables on the statistical moments ($\mu$, $\sigma$) of the bearing capacity and on Sobol indices.

<table>
<thead>
<tr>
<th>Coefficient of variation</th>
<th>$\mu$</th>
<th>$\sigma$</th>
<th>SU(GSI)</th>
<th>SU($m_i$)</th>
<th>SU($\sigma_c$)</th>
<th>SU(D)</th>
</tr>
</thead>
<tbody>
<tr>
<td>COV(GSI)</td>
<td>5%</td>
<td>1.4937</td>
<td>0.4396</td>
<td>0.1031</td>
<td>0.1165</td>
<td>0.7207</td>
</tr>
<tr>
<td></td>
<td>10%</td>
<td>1.5056</td>
<td>0.5129</td>
<td>0.3141</td>
<td>0.0866</td>
<td>0.5378</td>
</tr>
<tr>
<td></td>
<td>15%</td>
<td>1.5254</td>
<td>0.6258</td>
<td>0.5063</td>
<td>0.0594</td>
<td>0.3712</td>
</tr>
<tr>
<td>COV($m_i$)</td>
<td>6.25%</td>
<td>1.5059</td>
<td>0.4948</td>
<td>0.3384</td>
<td>0.0233</td>
<td>0.5789</td>
</tr>
<tr>
<td></td>
<td>12.5%</td>
<td>1.5056</td>
<td>0.5129</td>
<td>0.3141</td>
<td>0.0866</td>
<td>0.5378</td>
</tr>
<tr>
<td></td>
<td>18.75%</td>
<td>1.5050</td>
<td>0.5426</td>
<td>0.2805</td>
<td>0.1743</td>
<td>0.4807</td>
</tr>
<tr>
<td>COV($\sigma_c$)</td>
<td>12.5%</td>
<td>1.5052</td>
<td>0.3895</td>
<td>0.5451</td>
<td>0.1502</td>
<td>0.2333</td>
</tr>
<tr>
<td></td>
<td>25%</td>
<td>1.5056</td>
<td>0.5129</td>
<td>0.3141</td>
<td>0.0866</td>
<td>0.5378</td>
</tr>
<tr>
<td></td>
<td>37.5%</td>
<td>1.5051</td>
<td>0.6700</td>
<td>0.1840</td>
<td>0.0507</td>
<td>0.7093</td>
</tr>
<tr>
<td>COV(D)</td>
<td>5%</td>
<td>1.4937</td>
<td>0.5062</td>
<td>0.3225</td>
<td>0.0888</td>
<td>0.5515</td>
</tr>
<tr>
<td></td>
<td>10%</td>
<td>1.5056</td>
<td>0.5129</td>
<td>0.3141</td>
<td>0.0866</td>
<td>0.5378</td>
</tr>
<tr>
<td></td>
<td>15%</td>
<td>1.5254</td>
<td>0.5239</td>
<td>0.3013</td>
<td>0.0832</td>
<td>0.5168</td>
</tr>
</tbody>
</table>

Effect of the correlation coefficient and the distribution type of the random variables

The PDFs presented in Figure 3 show that the increase in the coefficient of correlation between GSI and $\sigma_c$ significantly increases the variability of the ultimate bearing capacity. The smallest response variability is thus obtained for negatively correlated variables.

Concerning the distribution type of the input random variables, two cases of normal and nonnormal (i.e. log-normal) random variables combined with two configurations of COVs were considered (Figure 4).

Figure 3. Influence of $\rho$(GSI, $\sigma_c$) on the PDF of the bearing capacity.

Figure 4. Influence of the distribution type of the input random variables on the PDF of the bearing capacity for two different sets of COVs.
The “standard COVs” corresponds to the reference case presented in Table 1 while the “high COVs” corresponds to these values increased by 50%. For these two sets of COVs, the non-normality of the distribution type of the input random variables has significant influence on the shape of the PDF of the ultimate bearing capacity as can be seen from Figure 4; the influence being more important for the case of “high COVs” than the case of “standard COVs”. Notice however that the probability distribution of the random variables has practically no influence on the mean and the standard deviation of the system response and consequently, no effect on the coefficient of variation of this response (results not shown here).

CASE OF AN INCLINED LOAD

Figure 5 shows the PDFs of the ultimate bearing capacity for different values of the load inclination $\alpha$. One can see that the variability of the ultimate bearing capacity is significant in the case of small load inclination $\alpha$ and it decreases when $\alpha$ increases. This may be explained by the fact that the response involved in the analysis is the ultimate bearing capacity. Thus, it would be logical to have the most significant variability when the punching is most predominant (i.e. for small load inclination).

![Figure 5. PDF of the bearing capacity for different cases of load inclination.](image)

Table 3. Statistical moments of the ultimate bearing capacity for different cases of load inclination $\alpha$.

<table>
<thead>
<tr>
<th>$\alpha$ (°)</th>
<th>5</th>
<th>10</th>
<th>15</th>
<th>23</th>
<th>30</th>
<th>35</th>
<th>40</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\mu$</td>
<td>1.3032</td>
<td>1.1004</td>
<td>0.9090</td>
<td>0.6367</td>
<td>0.4405</td>
<td>0.3260</td>
<td>0.2325</td>
</tr>
<tr>
<td>$\sigma$</td>
<td>0.4443</td>
<td>0.3758</td>
<td>0.3104</td>
<td>0.2178</td>
<td>0.1509</td>
<td>0.1119</td>
<td>0.0800</td>
</tr>
<tr>
<td>COV</td>
<td>0.3410</td>
<td>0.3415</td>
<td>0.3415</td>
<td>0.3421</td>
<td>0.3426</td>
<td>0.3432</td>
<td>0.3439</td>
</tr>
<tr>
<td>$\delta$</td>
<td>1.0627</td>
<td>1.0672</td>
<td>1.0693</td>
<td>1.0698</td>
<td>1.0709</td>
<td>1.0712</td>
<td>1.0742</td>
</tr>
<tr>
<td>$\kappa$</td>
<td>2.0710</td>
<td>2.0928</td>
<td>2.1113</td>
<td>2.0969</td>
<td>2.1051</td>
<td>2.095</td>
<td>2.1153</td>
</tr>
</tbody>
</table>

Table 3 shows that the mean and standard deviation of the ultimate bearing capacity vary with the load inclination. However, the coefficient of variation, the skewness $\delta$ and the kurtosis $\kappa$ do not vary with this inclination; this observation may be explained by the fact that the variability of the input random variables is similar for the different load inclinations.
CONCLUSION

A probabilistic analysis of a centrally loaded strip foundation resting on a rock mass obeying Hoek-Brown failure criterion has been performed. The deterministic model was based on the kinematical approach of the limit analysis theory. The polynomial chaos expansion methodology was used for the probabilistic analysis. The uncertain parameters involved in the analysis were the Geological Strength Index ($GSI$), the uniaxial compressive strength of the intact rock ($\sigma_c$), the intact rock material constant ($m_i$) and the disturbance coefficient ($D$). For the vertically loaded footing, it was shown that (i) a fourth order PCE is sufficient to accurately represent the system response; (ii) the parameters $m_i$ and $D$ have a smaller effect on the variability of the ultimate bearing capacity compared to the other two parameters $GSI$ and $\sigma_c$; (iii) the assumption of negatively correlated variables leads to less spread out PDF of the ultimate bearing capacity; (iv) the non-normality of the input random variables has a significant impact on the shape of the PDF of the ultimate bearing capacity, there is an almost no effect on its mean and standard deviation; (v) the COV of the random variables has practically no effect on the mean value of the ultimate bearing capacity, the increase in the COV of the parameters $GSI$ or $\sigma_c$ has a significant effect on the variability of the ultimate bearing capacity; (vi) the increase in the COV of a certain variable increases its Sobol index (i.e., it increases its weight in the variability of the system response) and decreases the Sobol indices of the other variables. For the inclined load case, it was found that the variability of the ultimate bearing capacity decreases with the increase of the footing load inclination; however the coefficient of variation of this bearing capacity is constant regardless of the load inclination.

REFERENCES