Introduction

Landsliding is a major issue faced by people throughout the world. It causes severe damage not only to properties, roads and buildings but a large number of people lost their lives due to this issue. Therefore experts of this field always trying to know about the main reasons behind landsliding. There are two cases in slope stability analysis.

1. Estimate the stability of any slope before it fails.

2. Estimate the stability after the slope fails.

If any slope is expected to fail, it must be analyzed that how much it is safe? And must be treated so that it remains stable. If it is not treated on time, it may result in huge loss of property as well as human lives like it happened in the past many times. Fig. 1 shows few glimpses of landsliding.

Normally geotechnical engineers are facing major problems in slope stability projects to know about the slope factor of safety. The soil behavior depends on many parameters, i.e. cohesion, friction, unit weight, moisture content, ground geology and temperature etc. Therefore, geotechnical engineers are always busy investigating of soil behavior, keeping all
the different parameters in considerations. In this regard, many researchers and investigators provided different techniques and formulas to predict the slope factor of safety in advance. An Artificial Neural Network was used to find out the correlations between different parameters of slope [1]. The results were compared with Hoek and Bray [2] which shows the results to be very satisfactory. To study the stability prediction of Letlhakane mine, some researchers also used the geomos slope monitoring system. The geomos slope monitoring system is a continuous and automatic system which runs for 24 hours. Evolutionary Polynomial Technique (EPR) was developed to predict the FS [3]. This model was very useful in predicting the behavior of slopes for analyzing FS. Later Geo Studio was used and compared with the fuzzy logic system [4,5]. The results were found very close to the target. In another work, as artificial neural network and multiple linear regression to calculate the FS in case of a typical artificial slope [6]. This slope was also subjected to seismic forces. The predicted results showed that the results were highly precise. Another researcher developed equations considering the depth ratio in case of rainfall induced slopes [7]. The results were compared with the previous steady state hydrological model and landslide inventory graphs, they concluded that the predicted approach gives very satisfactory results. Using the theory of mass approach, the run out distance of rotational type slope was predicted in other work [8]. They concluded that the unit weight plays a very important role in the factor of safety issues. In another analysis it was concluded that the Chauncun Landslide was triggered by the excavation of the slope toe and the landslide was a partial revival of an ancient landslide and was a thrust load-caused landslide [9]. Other researchers worked on same issue to determine the slope safety by considering different parameters which affect the slope [10-16].

Most of the previous work is based on slope stability analysis of already failed slope. In landsliding issues, it is required to know about the slope stability in advance. Or if any slope already exist and seems to be risky, it must be analyzed and treated such that the stability gets increased. Moreover the slope factor of safety varies with different slope angles (\( \theta \)). Fig. 2 shows the slope model considered in this research.

In this paper, prediction equations are developed using multiple linear regression. A software namely Statistical Package for the Social Sciences (SPSS) is used to find the correlations of FS and slope angle. This software is basically used for analyzing complex data and one of the main purpose is to make correlations between different variables. SPSS can do linear as well as non-linear analysis. Fig. 3 shows the flowchart of the methodology of this paper:
Material and Methods

Local Chinese site is considered for the material to be analyzed. After all the necessary experiments, such as Atterberg limits, moisture content test, triaxial test and all other necessary tests, details of the material properties achieved are given in Table 1. In this table, material 1 to 5 is pure clay while material 6 to 10 is clayey sand. The slope angle ($\beta$) range is 70 to 88 degrees. It is because the most critical slope angle is in range of 60 to 90 degrees.

Ten number of different material types were considered and for each material type, slope angle is changed from 70 to 88 degrees. Hence a total number of four hundred analysis are performed. All this analysis is performed in five phases.

- Phase 1: Unsaturated non-seismic analysis
- Phase 2: Saturated non-seismic analysis
- Phase 3: Unsaturated seismic analysis
- Phase 4: Saturated seismic analysis
- Phase 5: Case study – Landslide in Karachi, Pakistan

In all these phases, correlations between FS and $\beta$ are developed. The applicability of these equations is very useful as the FS have a very close relation with $\beta$.
Results and Discussion

Unsaturated Non-Seismic Analysis

Fig. 4 shows the slope model considered in phase 1. Table 2 shows the factor of safety values achieved in all the hundred analysis for different material types and varying slope angles.

Table 3 and 4 shows the correlation equations in case of clay and clayey sand respectively for all the ten types of materials in case of unsaturated non-seismic state.

From Table 3, the final average correlation between FS and $b$ in case of clay came out to be:

$$FS = 1.836 - 0.011*b$$  \hspace{1cm} (1)
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Also from Table 4, the final average correlation between FS and $\beta$ in case of clayey sand came out to be:

$$FS = 0.77 - 0.004*\beta$$  \hspace{1cm} (2)

The coefficient of determination $R^2$ in equation 1 and 2 is 99.5 % which shows the variables are best correlated with each other.

In most of the cases, the FS value is less than 1. And less than 1 means the slope is a failed slope. The FS...
Maximum Shear Strength Variation with Slope Angle

The shear strength of these materials can be characterized by the equation.

\[ SS = \sigma \tan \phi \]  

...where SS is the shear strength, \( \sigma \) the effective normal stress on the failure plane, and \( \phi \) the effective stress angle of internal friction. Measuring or estimating the drained strengths of these materials involves determining or estimating appropriate values of \( \phi \).

SS is one of the main parameters which are responsible for the stability of any slope. This parameter is directly or indirectly affected by many other parameters such as cohesion, friction, unit weight, pore water pressure, moisture content, particle size, and slope angle (\( \beta \)).

Fig. 5 shows the maximum shear strength graph in case of saturated non-seismic case for material 1 (M1). The maximum shear strength in this case for \( \beta \) equals 70 is 44.702 KPa. The distance in meter shows the horizontal distance from 0, 0 coordinate, that is point 1 in Fig. 2.

Table 5 shows the shear strength values achieved in all the hundred analyses for different material types and varying slope angles.

Table 6 shows the correlation equations in case of clay for all the ten types of materials in case of unsaturated non-seismic state.

From Table 7, the final average correlation between shear strength and \( \beta \) in case of clay came out to be:

\[ \text{Shear strength} = - 16.686 + 0.92*\beta \]  

The factor of safety, FS, is defined with respect to the shear strength of the soil as:

\[ F = \frac{s}{\tau} \]  

...where \( s \) is the available shear strength and \( \tau \) is the equilibrium shear stress. The equilibrium shear stress is the shear stress required to maintain a just-stable slope and from Eq. (5) may be expressed as:

\[ \tau = \frac{s}{F} \]  

The shear strength can be expressed by the Mohr–Coulomb equation. If the shear strength is expressed in terms of total stresses, Eq. (6) is written as:

\[ \tau = \frac{c + \sigma \tan \phi}{F} \]  

If the shear strength is expressed in terms of effective stresses (e.g., drained shear strengths are being used), the only change from the above is that Eq. (7) is written in terms of effective stresses as:

\[ \tau = \frac{c/+(\sigma-u)\tan \phi}{F} \]  

Hence the shear strength can be correlated to all these parameters. Similarly the slope angle (\( \beta \)) could be correlated to all these parameters too. The correlations

| \( \beta \) | M1 | M2 | M3 | M4 | M5 |
|---|---|---|---|---|---|
| 70 | 44.702 | 47.2 | 51 | 54.12 | 57.98 |
| 72 | 41.422 | 43.8 | 47.3 | 50.2 | 53.7 |
| 74 | 42.8 | 45.26 | 48.9 | 51.9 | 55.5 |
| 76 | 47.13 | 49.7 | 53.8 | 57.1 | 61.15 |
| 78 | 51.5 | 54.2 | 58.7 | 62.3 | 66.8 |
| 80 | 47.4 | 57.8 | 54 | 57.4 | 61.5 |
| 82 | 52.1 | 54.9 | 59.4 | 63 | 67.7 |
| 84 | 49.6 | 52.2 | 56.6 | 57.2 | 64.3 |
| 86 | 54.6 | 57.4 | 59.6 | 66.1 | 70.9 |
| 88 | 59.7 | 62.7 | 68 | 72.2 | 77.6 |

Table 6. Shear strength in case of unsaturated non-seismic analysis.
of relative density, cone resistance and angle of internal friction given in Table 8.
Where:

\[
D_r = \frac{e_{\text{max}} - e}{e_{\text{max}} - e_{\text{min}}} \times 100
\]  

(9)

Hence using the above table, the slope angle can be further correlated to angle of internal friction, cone resistance and relative density. Duncan [17] also provided the correlations of state of packing with blow count in Standard Penetration Test (SPT).

Saturated Non-Seismic Analysis

One of the major causes of landslide is rainfall. It is because the rain water increases the weight of the soil particles and hence decreases the shear strength [18-23].

In second phase of this work, the same slope is analyzed as unsaturated slope. Fig. 6 shows the saturated model. Table 9 shows the factor of safety values achieved in all the hundred analysis for different material properties and slope angles.

Table 9 shows the factor of safety values achieved in all the hundred analysis for different material properties and slope angles.

Table 10 and 11 show the correlation equations in case of clay and clayey sand respectively for all the ten types of materials in case of saturated state.

From Table 10, the final average correlation between FS and \( \beta \) in case of clay came out to be:

\[
FS = 1.302 - 0.008\beta
\]  

(10)

Also from Table 11, the final average correlation between FS and \( \beta \) in case of clayey sand came out to be:

Fig. 6. Slope model – saturated non-seismic case.
The coefficient of determination $R^2$ in equation 3 and 4 is 99.8 % which shows the variables are best correlated with each other.

Unsaturated Seismic Analysis

Another most critical situation is to find the slope stability in case of earthquake. Normally the horizontal seismic coefficients is considered in the analysis as the vertical seismic coefficient is almost negligible in most of the cases. Therefore in this phase, the seismic analysis is performed in which the horizontal seismic coefficient is considered as the maximum. That is 0.3 horizontal. Fig. 7 shows the slope model assumed in this case.

Table 12 shows the FS in case of unsaturated seismic analysis.

Table 13 and 14 shows the correlation equations in case of clay and clayey sand respectively for all the ten types of materials in case of unsaturated seismic analysis.

From Table 13, the final average correlation between FS and $\beta$ in case of clay came out to be:

$$FS = 1.22 - 0.007*\beta$$  \hspace{1cm} (12)

Also from Table 14, the final average correlation between FS and $\beta$ in case of clayey sand came out to be:

$$FS = 0.50 - 0.003*\beta$$  \hspace{1cm} (13)

The coefficient of determination $R^2$ in equation 5 and 6 is 99.3 % which shows the variables are best correlated with each other.

Saturated Seismic Analysis

Fig. 8 shows the slope model in this case.

Table 15 shows the FS in case of saturated seismic analysis.

Table 9. FS in case of saturated non-seismic analysis.

| $\beta$ | M1  | M2  | M3  | M4  | M5  | M6  | M7  | M8  | M9  | M10 |
|---|---|---|---|---|---|---|---|---|---|---|
| 70 | 0.674 | 0.728 | 0.757 | 0.792 | 0.818 | 0.174 | 0.230 | 0.279 | 0.324 | 0.377 |
| 72 | 0.656 | 0.709 | 0.736 | 0.769 | 0.794 | 0.168 | 0.224 | 0.272 | 0.314 | 0.367 |
| 74 | 0.645 | 0.696 | 0.723 | 0.755 | 0.778 | 0.162 | 0.218 | 0.266 | 0.306 | 0.356 |
| 76 | 0.628 | 0.680 | 0.704 | 0.736 | 0.757 | 0.158 | 0.211 | 0.259 | 0.301 | 0.348 |
| 78 | 0.614 | 0.663 | 0.688 | 0.719 | 0.740 | 0.156 | 0.204 | 0.253 | 0.293 | 0.341 |
| 80 | 0.597 | 0.645 | 0.670 | 0.701 | 0.724 | 0.152 | 0.200 | 0.245 | 0.286 | 0.334 |
| 82 | 0.587 | 0.633 | 0.659 | 0.689 | 0.712 | 0.149 | 0.197 | 0.239 | 0.278 | 0.326 |
| 84 | 0.573 | 0.620 | 0.643 | 0.673 | 0.694 | 0.147 | 0.195 | 0.235 | 0.271 | 0.317 |
| 86 | 0.559 | 0.605 | 0.627 | 0.655 | 0.675 | 0.144 | 0.191 | 0.232 | 0.267 | 0.310 |
| 88 | 0.546 | 0.590 | 0.612 | 0.640 | 0.661 | 0.143 | 0.189 | 0.229 | 0.263 | 0.305 |

Table 10. Correlations in case of Clay – saturated non-seismic case.

| Clay | M1 | M2 | M3 | M4 | M5 |
|---|---|---|---|---|---|
| FS | 1.166-0.007$*\beta$ | 1.258-0.008$*\beta$ | 1.309-0.008$*\beta$ | 1.368-0.008$*\beta$ | 1.410-0.009$*\beta$ |

Table 11. Correlations in case of Clayey Sand – saturated non-seismic case.

| Clayey Sand | M6 | M7 | M8 | M9 | M10 |
|---|---|---|---|---|---|
| FS | 0.288-0.002$*\beta$ | 0.388-0.002$*\beta$ | 0.479-0.003$*\beta$ | 0.561-0.003$*\beta$ | 0.653-0.004$*\beta$ |
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Fig. 7. Slope model – unsaturated seismic case.

Table 12. FS in case of unsaturated seismic analysis.

| β  | M1  | M2  | M3  | M4  | M5  | M6  | M7  | M8  | M9  | M10 |
|----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 70 | 0.669 | 0.724 | 0.734 | 0.759 | 0.769 | 0.155 | 0.218 | 0.273 | 0.319 | 0.372 |
| 72 | 0.652 | 0.704 | 0.714 | 0.740 | 0.750 | 0.148 | 0.207 | 0.262 | 0.307 | 0.360 |
| 74 | 0.640 | 0.693 | 0.702 | 0.727 | 0.735 | 0.137 | 0.198 | 0.250 | 0.296 | 0.349 |
| 76 | 0.623 | 0.674 | 0.683 | 0.707 | 0.716 | 0.131 | 0.194 | 0.241 | 0.284 | 0.338 |
| 78 | 0.608 | 0.657 | 0.666 | 0.690 | 0.698 | 0.123 | 0.184 | 0.236 | 0.275 | 0.325 |
| 80 | 0.595 | 0.644 | 0.652 | 0.675 | 0.684 | 0.117 | 0.180 | 0.229 | 0.270 | 0.317 |
| 82 | 0.586 | 0.633 | 0.642 | 0.665 | 0.673 | 0.113 | 0.173 | 0.223 | 0.264 | 0.312 |
| 84 | 0.572 | 0.620 | 0.628 | 0.650 | 0.658 | 0.110 | 0.168 | 0.218 | 0.258 | 0.306 |
| 86 | 0.560 | 0.604 | 0.613 | 0.635 | 0.644 | 0.107 | 0.166 | 0.213 | 0.254 | 0.300 |
| 88 | 0.553 | 0.597 | 0.606 | 0.627 | 0.636 | 0.104 | 0.164 | 0.210 | 0.248 | 0.296 |

Table 13. Correlations in case of clay – unsaturated seismic case.

| Clay | M1 | M2 | M3 | M4 | M5 |
|------|----|----|----|----|----|
| FS   | 1.121-0.007*β | 1.216-0.007*β | 1.230-0.007*β | 1.274-0.007*β | 1.287-0.007*β |

Table 14. Correlations in case of clayey sand – unsaturated seismic case.

| Clayey Sand | M6 | M7 | M8 | M9 | M10 |
|-------------|----|----|----|----|-----|
| FS          | 0.350-0.003*β | 0.422-0.003*β | 0.506-0.003*β | 0.580-0.004*β | 0.664-0.004*β |
Fig. 8. Slope model – saturated seismic case.

Table 15. FS in case of saturated seismic analysis.

| Slope angle (β) | M1   | M2   | M3   | M4   | M5   | M6   | M7   | M8   | M9   | M10  |
|-----------------|------|------|------|------|------|------|------|------|------|------|
| 70              | 0.396| 0.433| 0.451| 0.474| 0.490| 0.030| 0.067| 0.104| 0.136| 0.173|
| 72              | 0.387| 0.422| 0.442| 0.465| 0.480| 0.027| 0.063| 0.098| 0.129| 0.167|
| 74              | 0.376| 0.410| 0.429| 0.451| 0.467| 0.024| 0.058| 0.093| 0.122| 0.159|
| 76              | 0.370| 0.403| 0.422| 0.444| 0.460| 0.022| 0.053| 0.088| 0.117| 0.152|
| 78              | 0.359| 0.392| 0.410| 0.431| 0.446| 0.020| 0.047| 0.083| 0.112| 0.146|
| 80              | 0.351| 0.385| 0.400| 0.419| 0.432| 0.018| 0.043| 0.078| 0.107| 0.142|
| 82              | 0.342| 0.375| 0.390| 0.409| 0.423| 0.016| 0.040| 0.073| 0.102| 0.136|
| 84              | 0.331| 0.363| 0.379| 0.398| 0.413| 0.014| 0.038| 0.069| 0.096| 0.131|
| 86              | 0.325| 0.355| 0.371| 0.390| 0.403| 0.013| 0.036| 0.066| 0.092| 0.126|
| 88              | 0.319| 0.349| 0.365| 0.383| 0.394| 0.012| 0.034| 0.064| 0.089| 0.121|

Table 16. Correlations in case of clay – saturated seismic case.

| Clay | M1                  | M2                  | M3                  | M4                  | M5                  |
|------|---------------------|---------------------|---------------------|---------------------|---------------------|
| FS   | 0.701-0.004*β       | 0.760-0.005*β       | 0.795-0.005*β       | 0.840-0.005*β       | 0.871-0.005*β       |
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Table 17. Correlations in case of clayey sand – saturated seismic case.

| Clayey Sand | M6   | M7   | M8   | M9   | M10  |
|-------------|------|------|------|------|------|
| FS          | 0.099-0.001*β | 0.198-0.002*β | 0.262-0.002*β | 0.317-0.003*β | 0.372-0.003*β |

Tables 16 and 17 shows the correlation equations in case of clay and clayey sand respectively for all the ten types of materials in case of saturated seismic analysis.

From Table 16, the final average correlation between FS and β in case of clay came out to be:

FS = 0.793 – 0.005*β  \hspace{1cm} (14)

Also from Table 17, the final average correlation between FS and β in case of clayey sand came out to be:

FS = 0.25 – 0.002*β  \hspace{1cm} (15)

The coefficient of determination R² in equation (7) and (8) is 99.7 % which shows the variables are best correlated with each other.

A very similar work is also done to develop correlations [24]. These correlations are very useful to apply when the factor of safety, shear strength or any other value such as cohesion, friction or unit weight etc. is required to geotechnical engineers. The results of simple reliability analyses are neither more accurate nor less accurate than factors of safety calculated using the same types of data, judgments, and approximations.

Although neither deterministic nor reliability analyses are precise, they both have value, and each enhances the value of the other. The simple types of reliability analyses described in this paper require only modest extra effort compared to that required to calculate correlations of factors of safety with all other parameters and hence they can add considerable value to the results of slope stability analyses.

Phase 5: Case Study – Landslide in Karachi, Pakistan

Heavy rainfall started in Sindh and Baluchistan from the 6 August and continued till the 7 August 2020 with intermissions. Continuous rain over a period of 24 hours caused massive flooding in Karachi, Hyderabad, Shaheed Benazirabad and Dadu of Sindh province. However, Tehsil Johi in Dadu district is the area which is greatly affected by flash floods. It has been reported that floods are not only damaging infrastructures and houses but also destroyed crops in Johi Tehsil. Government of Sindh has declared 80 villages in Dadu district as “Calamity Affected Areas” [25]. The affected areas can be seen in Fig. 9.
During this flood, a landslide occurs at the Gulistan-e-Johar city of Karachi. This landslide damaged 22 to 30 cars as they were standing at the bottom of parking lot near the landslide hill. The site of this landslide was basically serving as parking lot to the residents of the area. Fortunately no death or injury was reported due to this vary landslide. A video of this landslide is also recorded by a nearby resident and is available online. Figs 10-13 show some different views of the landslide.

Apart from the landslide, the floods also caused a huge damage to crops and affected many people. Severity of the flood can be seen in Figs 14 and 15:

Fig. 10. Gulistan-e-Johar Landslide 2020 view which show three slipped steps.

Fig. 11. Gulistan-e-Johar Landslide 2020 view showing one of the damaged car.

Fig. 12. Gulistan-e-Johar Landslide 2020 view from front side.

Fig. 13. Gulistan-e-Johar Landslide 2020 concrete block fallen down.

Fig. 14. A police van stuck in water during the flood August 2020 at Karachi.

Fig. 15. Car stuck in water during the flood August 2020 at Karachi.
In one of the report [26] by European Commission’s Directorate-General for European Civil Protection and Humanitarian Aid Operations says:
- Monsoon rains and associated flooding continue to affect Pakistan, resulting in at least 163 fatalities, and more than 100 injured, as reported by national authorities on 30 August. More than 1,590 houses, nine bridges and 10 roads were damaged or destroyed.
- Over 20-29 August, heavy rain and urban flooding occurred in Karachi City (Sindh Province), leading to at least 27 fatalities and 13 injured. National authorities are carrying out rescue and relief activities, while food items have been distributed to the affected population.
- Heavy rain is forecast over most of Punjab, northern Khyber Pakhtunkhwa, Gilgit-Baltistan, and AJK Regions on 31 August - 1 September.

After collecting soil samples from the landslide site and tested in soil laboratory, the material properties were found similar as mentioned in table 1. The failed slope is modelled in Slide and back analysis is performed. Applying the correlations mentioned in equations 1, 2, 10, 11, 12, 13, 14 and 15, it is concluded that the slope factor of safety at the time of failure was less than 1. All such landslides can be controlled by analyzing the slope on time and fix it by either stepping technique or nailing. Figs 16 and 17 shows stepping and nailing technique.

The equations developed in this paper can contribute in it. This type of evaluation was not made in past because only FS were computed to guide the design. Now the FS value can be compared with all other parameters using all these correlations developed in this paper. This research is applicable for clay and clayey sand only. For all other soil types, it is recommended to develop new correlations.

**Conclusions**

The following conclusions are drawn from this research work:
1. The slope angle and factor of safety have a closure relationship with each other. More the slope angle, less will be the factor of safety. This variation can be observed in equations 1, 2, 10, 11, 12, 13, 14 and 15.
2. These correlations are the main outcome of this work. And they can be used to find the value of FS with any varying slope angle ($\beta$). By using these equations, a geotechnical engineer can calculate the optimum FS value while designing any slope.
3. Engineers can apply these correlations in designing an earthen dam or slope designing in mountainous regions while building a road adjacent to the mountain provided that the material properties and slope geometries are similar as considered in this paper.
4. These equations are applicable for homogenous slopes only. The material properties must be in the range of Table 1.
5. Future work can be done in case of non-homogenous slopes as well as complex shape slopes to get the correlations between different soil parameters, FS and $\beta$.
6. $\tau = c + \tan \phi$, using this equation, the correlations between shear strength and shear stress can be calculated in any future work.
7. Severe rainfall causes floods especially in urban and developed areas and ultimately it damages property and causes great human loss. One of such incident happened in Gulistan-e-Johar area of Karachi city on August 24, 2020. The landslide damages 22 to 30 cars and fortunately no human death or injury was reported. The correlations from equation 1 to 9 shows that the FS at the time of failure was less than 1. The slope angle must be kept at lower value in future to minimize such failure.

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Conflict of Interest

The authors declare no conflict of interest.

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