Empirical relationships of shear wave velocity, SPT-N value and vertical effective stress for different soils in Mashhad, Iran

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ABSTRACT

Shear wave velocity, V_s, is one of the important input parameters in seismic response analysis of the ground. Various methods have been examined to measure the soil V_s directly. Direct measurement of V_s is time consuming and costly, therefore many researchers have been trying to update empirical relationships between V_s and other geotechnical properties of soils such as SPT Blow count, SPT-N. In this study the existence of a statistical relationship between V_s , SPT- N_{60} and vertical effective stress, $\sigma_{\nu}{}'$, is investigated. Data set we used in this study was gathered from geotechnical and geophysical investigations reports. The data have been extracted from more than 130 numbers of geotechnical boreholes from different parts of Mashhad city. In each borehole the V_s has been measured by downhole method at two meter intervals. The SPT test also has performed at the same depth. Finally relationships were developed by regression analysis for gravels, sands and fine grain soils. The proposed relationships indicate that V_s is strongly dependent on $\sigma_{\nu}{'}.$ In this paper the effect of fine percent also is considered on the V_s estimation.

1. Introduction

The first step in the dynamic response analysis of soil is recording the soil key properties such as density, stiffness, damping and maximum shear modulus. The maximum shear modulus of soil is function of the shear wave velocity, V_s , and soil density $G_{max} = \rho V_s^2$. The V_s can be determined directly by testing undisturbed soil samples in laboratory or computed using *in-situ* measurements [Brandenberg et al. 2010]. The V_s is also used for the seismic soil classification in building codes. Some of the building codes such as NEHRP, UBC and Iranian seismic code (Iranian Code of Practice for Seismic Resistant Design of Buildings, Standard No. 2800) use the average of V_s up to 30-m to classify sites for earthquake engineering design [Borcherdt 1994, Dobry et al. 2000, BHRC 2005].

There are various *in-situ* methods for measuring V_s including seismic refraction survey, Down-hole, cross-

hole, seismic CPT, refraction microtremor and the spectral-analysis-of-surface-waves (SASW). It is notable that resolutions of these methods are different and depend on measurement depth, soil condition, interpretation methods and implementation details [Andrus and Stokoe 1997].

The most appropriate methods for measuring V_s are *in-situ* methods. But these methods are usually time consuming, costly, noisy and not applicable at every places. For nearly four decades, researchers have been trying to develop empirical relationships between V_s and other geotechnical properties of soils. Previous studies considered different parameters, including geology condition, stratigraphy, geological age, depth, grain size distribution, density, fine percent, standard penetration test value (SPT-N), cone penetration test (CPT) and vertical effective stress, σ_{ν} , to develop empirical relationships [Imai and Tonouchi 1982, Jafari et al. 1997, Hasancebi and Ulusay 2007, Brandenberg et al. 2010, Ghorbani et al. 2012].

In the most relationships, V_s is considered as a function of SPT-N values. The SPT is one of the effective tests in quick and low cost assessment of the soil mechanical properties; hence it has received the greatest attention from both academic researchers and professional geotechnical engineers. The purpose of the present study is to assess the correlations between V_s and soil properties using statistical methods for collected data throughout the Mashhad City in northeast of Iran (Figure 1). The correlational relationships would be used to assess the site response and to prepare the V_s model of the Mashhad depositional basin.

2. Review of available empirical correlations

Among the ancient laboratory works, Hardin and Richart [1963] performed one of the first comprehen-

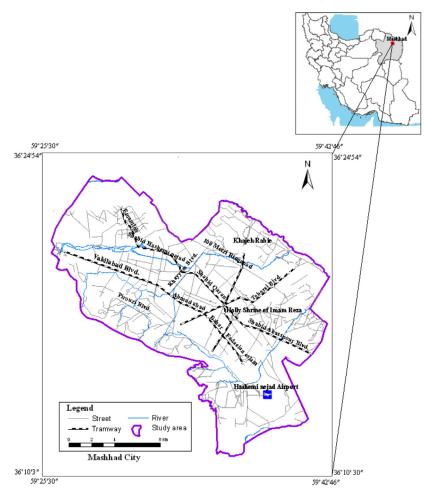


Figure 1. The location map of the study area.

sive laboratory studies in order to determine the V_s of different soils, and noted that the V_s values would be decreased by increasing the void ratio and would not be affected by the grain size distribution and relative density. Also they simulated the history of field loading condition on preloaded sands and found that the impact of load history on the V_s measurements was negligible. Hamilton [1971] assessed the effect of σ_{ν} on the V_s using laboratory tests and presented relations for coarse and fine sands. Lawrence [1965] presented V_s as a function of the one-fourth power of the effective hydrostatic stress, σ_0 .

Most of the previous studies have focused on the correlations between V_s and SPT-N values and just few studies have considered other parameters, including σ_{ν} , total stress, fine percent, depth and CPT. The great majority of correlations assessed in early efforts proposed a power law relationship between SPT-N and V_s as follow:

$$V_S = A.N^B \tag{1}$$

where V_s is shear wave velocity, N is SPT-N value, A and B are constants determined by statistical regression.

Early studies were often utilized the uncorrected SPT-N value to present relationships [Kanai 1966, Sakai

1968, Imai and Yoshimura 1970, Ohba and Toriumi 1970, Imai and Tonouchi 1982, Lee 1990, Iyisan 1996, Jafari et al. 1997, Jafari et al. 2002]. Pitilakis et al. [1999] suggested relationships between V_s and energy-corrected SPT-N for sands and clays by an energy correction factor of 75%. Hasancebi and Ulusay [2007] noted that the best relationships were achieved using uncorrected SPT-N values.

Some researchers have evaluated the impact of soil type on the correlations between V_s and SPT-N. Sykora and Stokoe [1983] assessed the influence of soil type on the mentioned correlations and noted that V_s are not affected by soil type. Iyisan [1996] also investigated the soil type effects on the correlation between V_s and SPT-N in an earthquake-prone area and proposed different relationships for gravels, sands and clays. Lee [1990] and Jafari et al. [2002] also presented different empirical relationships for clays and sands. Kokusho and Yoshida [1997] found that the correlations for gravely soils were dependent on the grain size distribution. Uma Maheswari et al. [2010] evaluated 200 data pairs and found that the effect of soil type is very low.

Campbell and Duke [1976] used 63 data pairs to develop the relationships for both old and recent alluvial soils. They found that the V_s values were dependent

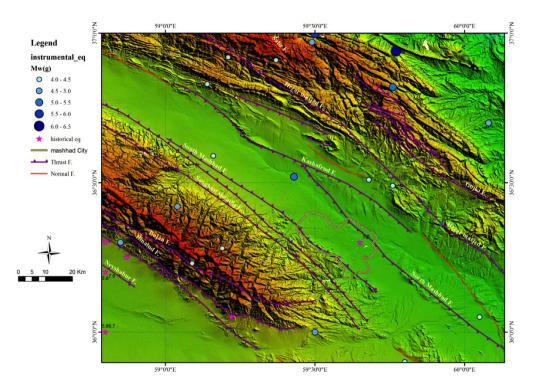


Figure 2. Seismotectonic map of Mashhad city.

only on the depth of soil layers. Sykora and Stokoe [1983] assessed the geological age and found it had no effect on the correlations. Ohta and Goto [1978] proposed a relationship which defines V_s as a function of SPT-N, depositional age, depth and lithofacies. Andrus et al. [2009] and Hayati and Andrus [2009] developed the relationships between V_s , SPT and CPT values for sands with different ages.

Shibata [1970] investigated the results of previous studies to obtain a relationship between V_s and SPT-N. He combined the various relationships and found that V_s was not dependent on both porosity and σ_{ν} '. Sykora and Stokoe [1983] evaluated the effect of total overburden stress, σ_{ν} , and σ_{ν} ' on the correlation of V_s , SPT-N and found that σ_{ν} ' was more effective than σ_{ν} . Brandenberg et al. [2010] developed the relationships which estimated V_s as a function of SPT-N and σ_{ν} ' for the sands, silts and clay soils. Akin et al. [2011] proposed the correlations between V_s , depth and SPT-N values for different type soils. Ghorbani et al. [2012] used previous published studies to compile a database including the V_s , σ_{ν} , and SPT-N values. They proposed multivariate relationships between V_s , corrected SPT-N and σ_{ν} '.

3. Geological setting and database

Mashhad city stands on the alluvial plain which is surrounded by the Kopet-Daq Mountains on the north and the Binaloud Mountains on the south. Geologically, the basement of the plain is composed of Neogene and pre-Neogene rocks in the central and northern parts, and granodiorite, ultrabasic, schist and phyllite rocks in

southern and west southern parts. The Quaternary fluvial and floodplain deposits above the basement rocks. Based on the deep boreholes and geophysical studies of Mashhad plain, the thickness of Quaternary deposits reaches to more than 250 m in the central part of the city.

Mashhad city is surrounded by active faults of Kashafrud, North Mashhad and South Mashhad, and is located in the high seismic hazard zone of Iran [BHRC 2005]. Sesimotectonic map of Mashhad city is presented in Figure 2.

The database contents include information of more than 130 boreholes with depth of 20 to 80 m. Geographical distribution of the boreholes in the study area is shown in Figure 3. V_s values were measured at 2 m intervals using downhole method.

Downhole seismic test is one of the appropriate tests for determining the in-situ profile of body wave velocity with depth. The test would be performed in single borehole with one or two geophone receivers. Standard test methods for downhole seismic testing are explained in ASTM D7400-08 [ASTM 2008]. In this study, the tests were performed using three-component geophone and RAS-24 digital seismograph. An impulse plank source was placed on the ground at 3 m away from the top of borehole and the single receiver swept the total length of borehole. The shear wave source was a 5 kg sledge hammer that horizontally hits both ends of wooden beam with steel end plate that was coupled to the ground by weighting it down with filled gunnysacks. The V_s was computed at distances of 2 m by measuring the distance and the travel time of the wave.

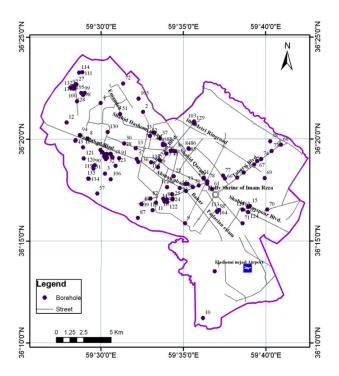


Figure 3. Location of geotechnical boreholes.

Other properties, including SPT-N value, soil texture, unit weight, natural moisture content, fine percent and Atterberg limits and groundwater table also were determined in the all boreholes. Properties of different types of soils are summarized in Table 1 by V_s class which was based on previous studies of authors [Ghazi et al. 2014]. Maximum values of N_{60} and unit weight, and minimum values of moisture content are observed in gravely soils. Geological logs with SPT and downhole test results for two boreholes are shown in Figure 4. For the statistical analyses of the present study, only the

cases were selected that SPT-N and V_s were recorded in the same depth. A total 1500 cases were used in statistical analysis. The frequency distribution of used data based on soil type is shown in Figure 5.

4. Data analysis and results

4.1. Correction of SPT-N value

SPT test results are affected by various factors such as equipments, σ_{ν} , and soil conditions. There are different equations for correction SPT-N values. In the present study we used McGregor and Duncan's equation that is the most comprehensive equation for the SPT-N correction [McGregor and Duncan 1998]:

$$N_{60} = (C_B C_C C_R C_{BF} C_S C_A C_E) N_{field}$$
 (2)

where C_B is borehole diameter correction factor, C_R is rod length correction factor, C_E is energy correction factor, C_S is liner correction factor, and C_C , C_{BF} and C_A are hammer cushion correction factor, blow count frequency correction factor and anvil correction factor, respectively. A borehole diameter correction factor of 1.0 was applied because in rotary drilling, the borehole diameter is <116 mm. C_R was applied based on the information in Table 2. Most of the consulting companies in Iran use a safety hammer with an efficiency of 60%, therefore C_E of 60% was applied. C_S of 1.0 was applied because the SPT sampler is not usually fit with the liner. C_C , C_{BF} and C_A were neglected because of the lack of efficient information.

In this study we use the common overburden cor-

V_s class	Soil	N_{60}	Yd	w	V_s class	Soil	N_{60}	Yd	w
<300	Clay	23±11	16.2±1	12.6±3	750-950	Clay	55±20	17.3±1	15.2±4
	Silt	16±7	16.9±1	8.8±5		Silt	45±13	17.4±1	18.4±4
	Sand	29±14	17.1±1	5.2±3		Sand	70±30	17.7±1	5.8±3
	Gravel	39±18	18.6±1	4.7±2		Gravel	75±39	19.1±1	5.4±3
300-500	Clay	26±15	16.5±1	13.4±4	>950	Clay	52±22	17.4±1	15.5±4
	Silt	27±12	17.6±1	12.9±3		Silt	-	-	-
	Sand	43±22	17.3±1	6.1±3		Sand	75±31	17.7±1	6±2
	Gravel	49±25	18±1	5.7±4		Gravel	77±39	19.4±2	5.4±3
500-750	Clay	39±23	16.8±1	14.2±5					
	Silt	38±22	16.2±1	17.4±6					
	Sand	56±27	17.4±1	5.8±3					
	Gravel	58±25	18.5±1	5.4±3					

Table 1. Median \pm standard deviation values of soil properties, V_s class is shear wave velocity in m/s N_{60} is SPT-N value corrected for energy in count per 30 centimeters, Yd is dry density in KN/m³ and w is moisture content in percent.

rection equations for correcting V_s and N_{60} that is mentioned in Equation (3) and Equation (4).

$$\left(N_{l}\right)_{60} = \left(\frac{P_{a}}{\sigma_{v}}\right)^{n} \cdot N_{60} \tag{3}$$

$$V_{SI} = \left(\frac{P_a}{\sigma_{,'}}\right)^m . V_S \tag{4}$$

where P_a is reference pressure and equals to 100 kPa, σ_{ν} ' is vertical effective stress and the exponents n and m are empirical constants that depend on soil type, cementation, and plasticity index. The exponents n and m are often assumed 0.5 and 0.25 respectively.

4.2. Statistical analysis

Equation (5) which is the rearrangement of Equations (1), (3) and (4) is used in this study:

$$\left(\frac{P_a}{\sigma_{,'}}\right)^m . V_S = A. \left(\frac{P_a}{\sigma_{,'}}\right)^n . N_{60}^B \tag{5}$$

By taking natural logarithms of Equation (5), the regression form of Equation (6) which is utilized in this study is obtained as:

$$LnV_{S} = \beta_{0} + \beta_{1}Ln(N_{60}) + \beta_{2}Ln(\frac{P_{a}}{\sigma_{..}}) + \varepsilon$$
 (6)

where V_S (average shear wave velocity) is in m/s, σ_{ν} is in kPa and N_{60} is in blow per 30 centimeters. β_0 , β_1 and β_2 are the regression constants, and ϵ is standard error.

The regression was performed using linear multiple regression algorithm in SPSS (Statistical Package for the Social Sciences). Linear Regression estimates the coefficients of the linear equation, involving one or more

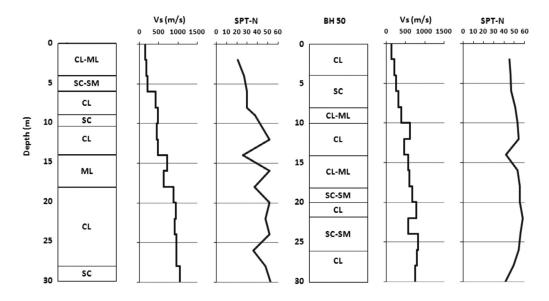


Figure 4. Geology, shear wave velocity and SPT-N profiles at two different sites (Soils are classified based on The Unified Soil Classification System. CL: Lean clay; CL-ML: Silty clay; ML: Lean silt; SC: Clayey sand; SC-SM: Silty, clayed sand).

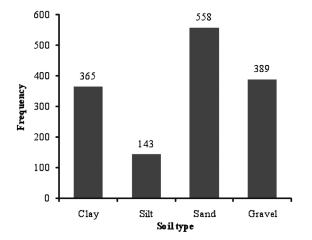


Figure 5. Frequency of data point based on soil type.

Rod length (m)	Correction factor
<3	0.75
3-4	0.8
4-6	0.85
6-10	0.95
>10	1

Table 2. Rod length correction factors [Youd et al. 2001].

independent variables that predict the best value of the dependent variable.

Using Equation (6) following regressions was obtained for gravels, sands and fine grain soils:

$$LnV_S = 5.43 + 0.192 Ln(N_{60}) - 0.246 Ln(\frac{P_a}{\sigma_{v}'})$$
 $R^2 = 0.748$ Gravely soils

$$LnV_S = 5.51 + 0.156Ln(N_{60}) - 0.279Ln(\frac{P_a}{\sigma_v'})$$
 $R^2 = 0.745$ Sandy soils (8)

$$LnV_S = 5.69 + 0.118Ln(N_{60}) - 0.242 \left(\frac{P_a}{\sigma_{\nu}'}\right)$$
 (9)
 $R^2 = 0.629$ Fine grain soils

Standard errors of the regressions for gravel, sand and fine grain soils are equal to 0.146, 0.170 and 0.151 respectively. $Ln(V_s)$ versus $Ln(N_{60})$ and $Ln(V_s)$ versus

 $Ln(\frac{P_e}{\sigma_v^{'}})$ are plotted for different soils and are shown in Figure 6. Median regression lines corresponding to the median, median plus and minus of $Ln(\frac{P_e}{\sigma_v^{'}})$ and $Ln(N_{60})$ are plotted through the data points.

There are so many methods for validation of the regression relationship. One of these methods is the residual analysis. Graphical residual plots including histogram of the residual, probability-probability (p-p) plot of regression standardized residual and studentized residual plot for gravel, sand and fine grain soils are shown in Figure 7. Histogram of the residual checks whether the variance is normally distributed. A symmetric bell-shaped histogram which is evenly distributed around zero indicates that the normality assumption is likely true as it seen in our plots. If residuals are not normally distributed, the model is violated. A P-P plot compares the empirical cumulative distribution function of a variable with a specified theoretical cumulative distribution function. This plot also shows whether the residuals or error terms are normally distributed or

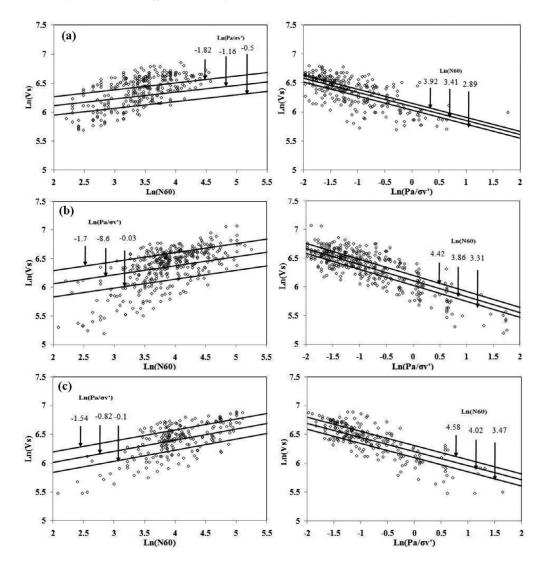


Figure 6. Graphical form of regression equations for (a) fine grain soil, (b) sandy soils and (c) gravely soils, With trend lines corresponding to the mean and mean plus and minus of $Ln(\frac{P_e}{\sigma'})$ and $Ln(N_{60})$.

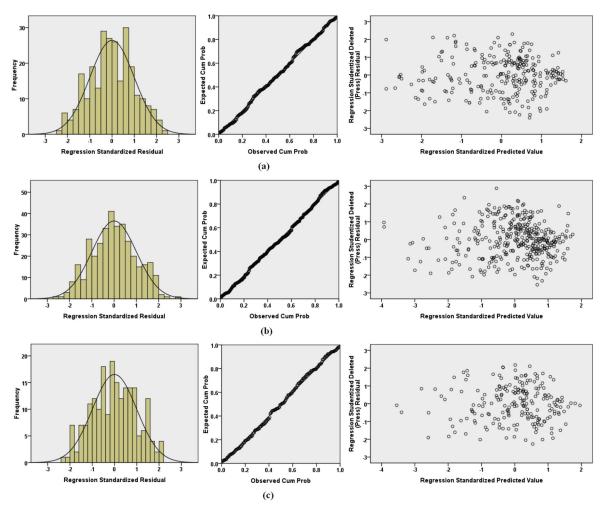


Figure 7. histogram of the residual, probability-probability (p-p) plot of regression standardized residual and studentized residual plot for (a) fine grain soil, (b) sandy soils and (c) gravely soils.

not. The points should be close to a diagonal line for normally distributed data. Therefore, the validity of presented regression analysis is approved. Also quality of the regression can be evaluated using residual plot. In this plot, the studentized deleted residual is plotted against the standardized estimated values of $V_{\rm s}$ variable. The standardized residual is the residual divided by its standard error. Standardizing is a method for transforming data so that, the mean is equal to zero and the standard deviation is equal to one. The studentized deleted residual is the residual that would be obtained if the regression was re-run omitting that observation from the analysis. For the good regression, the scatter plot of the residuals should be horizontal and random.

Measured V_s and estimated V_s are compared in Figure 8 for each correlation in order to assess the performance of the regression models. Data points are scattered between lines 1:0.65 and 1:1.5. Majority of the data points are plotted close to the line 1:1. The data points with V_s of less than 500 m/s are overestimated while the data points with V_s of more than 800 m/s are underestimated.

4.3. Influence of fine percent on V_s estimation

One of the effective parameters on SPT-N value is fine percent (passing No. 200), especially in coarse grain soils. In this study, the effect of fine percent was also evaluated. For this purpose, all data were classified by the fine percent regardless of soil type. The classification boundaries were considered based on the unified soil classification. At first all data were divided into the four categories of < 5, 5-12, 12-50 and more than 50. There was a few data into first class (<5), therefore two first classes were merged together and the class of <12% fine percent was considered. The regression models and adjusted R^2 for all three groups are presented in Table 3.

Results of the regressions are shown in Figure 9. Coarse soils show less V_s than fine soils in the low values of N-SPT. By increasing N-SPT, the V_s values of coarse soil increase further rather than fine soils. It is because of the inherent ability of fine soils in transferring the dynamical energy which is independent to the density and void ratio of soils. All curves were determined according to $\frac{P_s}{\sigma_s} = 1$. For estimating the real V_s , the obtained V_s values from the mentioned curves

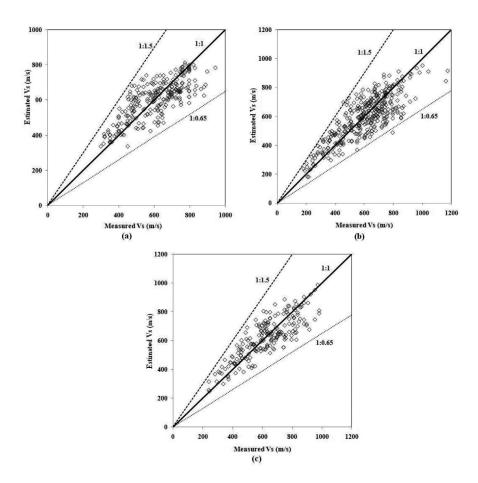


Figure 8. Measured versus estimated V_s for (a) fine grain soil, (b) sandy soils and (c) gravely soils with lines of 1:1.5, 1:1 and 1:0.65.

Adjusted R ²	Regression model	Soil groups (based on fine percent)
0.829	$LnV_S = 5.25 + 0.234 Ln(N_{60}) - 0.287 Ln(\frac{P_a}{\sigma_v})$	<12
0.701	$LnV_S = 5.52 + 0.159 Ln(N_{60}) - 0.257 Ln(\frac{P_a}{\sigma_{\nu}'})$	12-50
0.629	$LnV_S = 5.69 + 0.118 Ln(N_{60}) - 0.242 Ln(\frac{P_a}{\sigma_y^{\prime}})$	>50

Table 3. Regression models and adjusted R² for all groups that are divided based on fine percent, LnV_S is natural logarithm of shear wave velocity, $Ln(N_{60})$ natural logarithm of N-SPT corrected for energy, $\frac{P_a}{\sigma_v^2}$ is vertical effective stress factor.

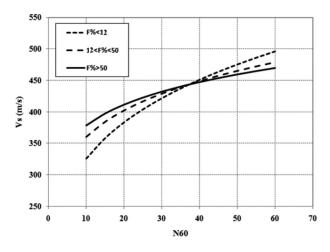


Figure 9. N_{60} and V_s relations for $\frac{P_s}{\sigma_r^2}=1$ as function of fine percent (F).

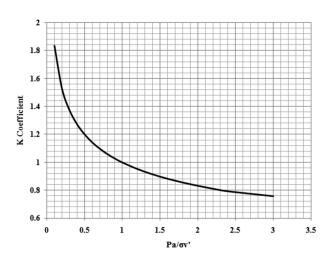


Figure 10. Curve of correction coefficient for determining real V_s for all kinds of soil.

must be multiplied by K coefficient which is shown in Figure 10.

5. Discussion

5.1. Relative impact of $\,N_{60}^{}$ and $\sigma_{\nu}^{'}$ on regression

When a multiple regression analysis carried out on variables, the software gives both unstandardized coefficients and standardized coefficients. The standardized coefficients are useful to find the contribution of each independent variable on estimating the dependent variable, when the variables are measured in different units of measurement. Standardization of each variable is done by software before regression analysis. For the purpose of this, the mean of variable is subtracted from each value of variable and then the difference value is divided by the standard deviation of the variable. The absolute coefficient value close to 1.0 shows a strong dependency. The standardized correlation coefficients of N_{60} and σ_{ν} for each soil categories are given in Table 4 in order to show their contribution on the V_s estimation. In all three models, the influence of σ_{ν} on estimating of V_s is stronger than the influence of N_{60} .

For assessing the influence of N_{60} and $\sigma_{\nu}{}'$ a simple

	Variable		
Soil category	$Ln(N_{60})$	$Ln\left(\frac{P_a}{\sigma_{\nu}'}\right)$	
Gravely soils	0.368	-0.610	
Sandy soils	0.257	-0.692	
Fine grain soils	0.239	-0.622	

Table 4. Standardized correlation coefficients of N_{60} and σ_{ν}' .

linear regression was also performed between LnV_S and LnN_{60} using the form of Equation (10) for all soil categories. The residuals of the regression model are plotted versus $Ln(\frac{P_s}{\sigma'})$ in Figure 11.

$$LnV_S = \alpha_0 + \alpha_1 Ln(N_{60}) + \varepsilon \tag{10}$$

The bias of residuals is clearly seen in the mentioned figure. The negative residuals are seen at the low values of σ_{ν} ' and the positive residuals are seen at the high values of σ_{ν} '. The negative residuals indicate the underestimation of V_s whereas the positive residuals indicate the overestimation of V_s . Therefore, neglecting of the σ_{ν} ' in the regression analysis can cause the significant error in the V_s estimation, implying the better V_s estimation using

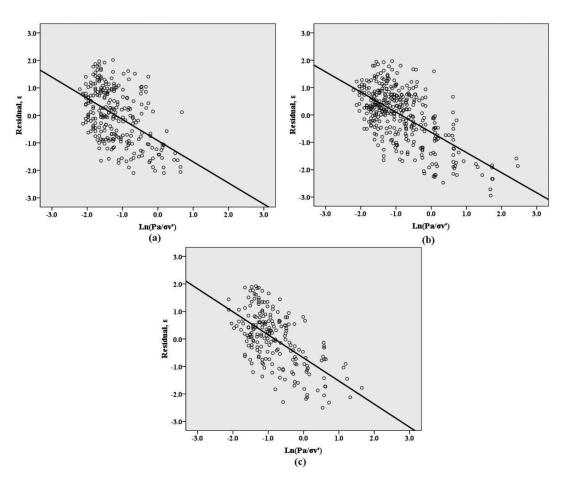


Figure 11. Residuals versus $Ln(\frac{P_a}{\sigma_v})$ for ordinary regression in order to show importance of effective overburden stress in V_s estimation for (a) fine grain soil, (b) sandy soils and (c) gravely soils.

 σ_{ν} alone rather than using N_{60} alone. Brandenberg et al. [2010], Akin et al. [2011] and Ghorbani et al. [2012] were also received to the similar findings.

5.2. Calculation of estimated V_s^{30}

One of the main applications of the correlations obtained in this study is the estimation of the upper 30 m V_s of the ground, V_s^{30} , which is the main parameter in some of the Seismic code as well as Iranian seismic code (Standard No. 2800). V_s^{30} is the most important parameter in Standard No. 2800 to classify the site conditions into the four categories (Table 5). For this, among the available databases, boreholes greater than 30 m in length (84 boreholes) were chosen to check the accuracy of V_s^{30} estimations. Both of measured V_s^{30} and estimated V_s^{30} for V_s profile measured directly and V_s profile estimated by Equation (7), Equation (8) and Equation (9) were computed from the time taken by the V_s to travel from depth of 30 m as:

$$V_s^{30} = 30 / \sum_i^n d / (V_s)_i \tag{11}$$

where d is the layer thickness or tributary length assigned to the estimated V_s and n is the number of layer or the number of V_s values estimated along the upper 30 m of the ground.

We compared the measured V_s^{30} and the estimated V_s^{30} in Figure 12. The data points are plotted between the lines of 1:0.75 and 1:1.3. As seen in Figure 12 most of the sites used in this study are located in the V_s^{30} category of 375-750 m/s which are classified as ground

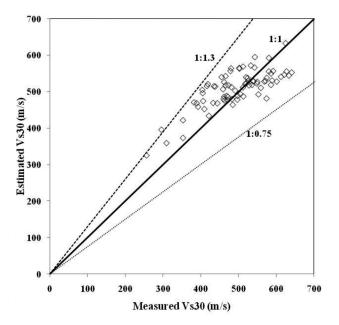


Figure 12. Estimated V_s^{30} versus measured V_s^{30} with lines of 1:1.3, 1:1 and 1:0.75 that shows good correlation between estimated and measured V_s^{30} .

type II according to the Standard No. 2800. Estimated V_s usually tend to higher V_s^{30} . According to measured V_s^{30} , just four sites were classified as ground type III while based on estimated V_s^{30} three of them were classified as ground type III and one site was classified as ground type II.

6. Conclusion

The shear wave velocity was estimated as a function of N-SPT value and effective overburden stress. Re-

$V_s^{30}(\text{m/s})$	Description	Ground type	
>750	a) Igneous rocks, stiff sedimentary rocks and massive metamorphic rocks and conglomerate.	I	
$375 \le V_s^{30} \le 750$	b) Stiff soils with a thickness more than 30 meters above bed rock.		
$375 \le V_s^{30} \le 750$	a) Loose and decomposed rocks due to weathering.	п	
$375 \le V_s^{30} \le 750$	b) Stiff soils having a thickness more than 30 meters above bed rock.	11	
$175 \le V_s^{30} \le 375$	a) Rocks that are disintegrated due to weathering.	III	
$175 \le V_s^{30} \le 375$	b) Layers of sand and gravel with medium compaction.	111	
<175	a) Soft deposits with high moisture content.	IV	
<175	b) Any soil profile containing clay with a thickness of more than 6 meters.	1 V	

Table 5. Soil profile classification according to Iranian seismic code of practice (Standard No. 2800). V_s^{30} is V_s averaged over the top 30 m of the ground.

gression relationships were obtained based on these parameters for gravel, sand and fine grain soils. The database has been compiled from 130 boreholes and totally 1500 data cases. Existing relationships around the world were reviewed which proposed the V_s as a function of different parameters. The majority of previous studies evaluated the V_s as a function of SPT-N alone. Most of the recent studies have investigated the effects of other parameters such as type soils, total overburden stress, effective overburden stress, depth and depositional age on the prediction of V_s . The results of this study indicate the neglecting of σ_{ν} in V_{ς} estimation would be tended to the significant errors. In this paper, the impact of fine percent was assessed regardless of soil types, and three curves were determined for graphical estimating of V_s . Obtained V_s from these curves should be multiplied by K coefficient which differs by variation of $\frac{P_a}{\sigma_a}$ ratios.

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