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Regression Modelling of California Bearing Ratio (CBR) Predicted from Index Ehiorobo, J. O. *Received: 11 December 2018 Accepted: 31 December 2018 Published: 15 January 2019*

6 Abstract

Obtaining California Bearing Ratios (CBR) of soils for road construction projects could be a
time-consuming and costly exercise. In order to reduce the time and cost of obtaining CBR
values of soils, this paper presents a mathematical relationship between index properties of
lateritic soils, which can easily be obtained from simple laboratory investigations, and their
CBR (soaked and unsoaked) values.

12

13 Index terms— california bearing ratio, index properties, multivariate regression model,

14 **1** Introduction

n highway design, bearing capacity of sub grade soil is of great importance in the determination of pavement 15 thickness (Forkenbrock and Weisbrod, 2001). The sub-grade layer, which is the bottommost layer, is mostly 16 17 affected as load comes upon it (Forkenbrock and Weisbrod, 2001). In Nigeria, California Bearing Capacity 18 (CBR) test is one of the most common and comprehensive method currently used to determine the sub-grade 19 strength. It is essentially a measure of the shear strength of a material at a known density and moisture content. The shear strength of soils can generally be considered in terms of Coulomb's Law, as discussed by Croney, 20 21 (1977). Subgrade plays an important role in imparting structural stability to the pavement structure as it receives loads imposed upon it by road traffic (Croney, 1977; Forkenbrock and Weisbrod, 2001). A range of factors influence 22 the CBR of a particular material. Carter and Bentley (1991) mentioned the soil type, density, moisture content 23 and method of sample preparation as playing important roles. Apart from the material properties themselves, 24 moisture conditions are also pivotal. The moisture conditions at which the material is to be used vary according 25 to climatic region, and as such the soaked CBR test is used to simulate the worst likely conditions in service and 26 27 the un-soaked simulate the normal field condition (Kumar, 2014). For determining soaked value of the CBR, the 28 sample is submerged in water for 96 hours prior to performing the penetration test.

In the tropics, lateritic soils are used as a road making material and they form the sub-grade of most tropical roads (Alayaki, 2012). Lateritic soil is generally believed to be a very good sub-grade material for road construction. Nigerian roads and highways are usually constructed on compacted lateritic soils foundation. Although some lateritic soils (especially gravelly aggregates) have been found to be quite good in pavement construction particularly those with appropriate geotechnical characteristics, the limited availability of these materials in the country is a challenge to constructing durable roads and highways (Alayaki ,2012).

A good highway or road is a gateway to national development as they create access to infrastructure (Okovido 35 and Musa, 2004). In Nigeria, the failure of engineering facilities such as roads and embankments has attracted 36 numerous opinions on the causes (Orie and Nweni, 2015). These failures have necessitated the need for research 37 38 which revealed that the causes of the highway failure were traceable to indiscriminate dumping of waste, the use of 39 substandard materials and incompetent contractors. Apart from these mentioned causes, insufficient knowledge 40 of the sub-grade of the intended site before use is also a contributing factor of failure (Orie and Nweni, 2015). 41 Huge amounts of money are spent on road maintenance on annual basis, yet the pavement does not last for a long period of time before its fails as a result of not knowing the condition of the sub-grade before design (Alayaki, 42 2012). CBR test is one of those parameters that serves as an indication of sub-grade soil strength and hence 43 the service-life of a pavement depends on the sub-grade (Sathawara and Patel, 2013). Comparing soaked and 44 un-soaked CBR will help to know the behavior of the soil before and after construction. Knowing this will help 45 to minimize the high rate of pavement failure, and money spent on yearly maintenance will be used for other 46

7 A) LABORATORY TESTS RESULTS

⁴⁷ projects that will boost the economic and social development of the country (Orie and Nweni, 2015;Alayaki, ⁴⁸ 2012).

The aim of this study is to develop a relationship between the index properties of lateritic soils and their soaked and unsoaked CBRs of lateritic soils. This relationship will help in quick assessment of CBRs of soils during the design stages of engineering projects.

⁵² 2 II. Materials and Method

⁵³ 3 a) Study Area

The study area covers Ebhohimi, and Ekpoma in Edo central senatorial zone of Edo state, Nigeria as shown in figure ??.

⁵⁶ 4 Figure 1: Location Map for the study area b) Sample ⁵⁷ Collection

In order to have sufficient and reliable data for the targeted analysis, soil samples were collected from the study area. The samples were collected along the road, and borrow pits. A total of Twenty (20) disturbed samples were collected, using hand auger at a depth of 1 m to 2 m. Some were taken from both side of the road within a reasonable sampling interval of 2 to 3 km. The sample locations are shown in Table 1.

⁶² 5 d) Analysis of Data using Multivariate Regression

To find the dependence of the measured geotechnical parameters on the soaked and un-soaked CBR, mathematical modeling using multivariate regression analysis was done (Bello, 2012). CBR values were taken as dependent variables and index properties (LL, PL, PI, OMC, MDD (compaction tests values), % passing of 0.075mm and 0.425mm) as independent variables.

Where y is the dependent variables, 1. The first phase in the model development was the transformation of the independent variables (%passing 0.075mm and 0.425mm sieve, liquid limit, plastic limit, plastic index, maximum dry density (g/cm 3), optimum moisture content) into readable codes that can be used as input files for the analysis.? ? ? ? ? , , , ,

73 2. The second phase was to define the dependent variables (Soaked and un-soaked CBR) and the model analysis 74 method. In this case, least square regression based on multivariate model was selected. 3. The third and final 75 phase was to compute the coefficient statistics, and assess the model strength using coefficient of determination, 76 thereafter generate the multivariate equations.

77 III.

78 6 Results and Discussion

⁷⁹ 7 a) Laboratory Tests Results

The results of the laboratory tests for Ebhohimi borrow pits are shown in Table 2. Based on the obtained test results from Ebhohimi borrow pit (Table 2), the soil is classified as A-7-5 (sandy soil). From the conventional Atterberg limit tests, liquid limit values are in the range of 46.31 to 58.17, plasticity limit values are of 15.5 to 26.45 and plasticity index value of 26.12 to 41.06 as shown in Table 2. Soils with liquid limit less than 30% are considered to be of low plasticity, those with liquid limit between 30% and 50% exhibit medium plasticity and those with liquid limit greater than 50% exhibit high plasticity (Arora, 2004). All samples exhibited high plasticity except sample 4 in pits 1 and 2 which exhibited medium plasticity.

The particle size distribution passing through 0.075mm and 0.425mm ranged between 38.2 to 52.16 and 61.08 87 to 73.35, which indicate fine grained soils, the soil can be classified as sandy soil (Arora, 2004). The unsoaked 88 CBR values ranged between 0.71 and 9.72, while its corresponding soaked samples range between 0.63 and 8.18%. 89 The percentage decreases from soaked CBR to unsoaked CBR. This implies that as water is absorbed into the 90 compacted specimen, the resistance to penetration becomes drastically reduced. It has been recommended by 91 Federal Ministry of Works and Housing that the values of CBR for road base, sub base and subgrade should 92 93 not be less than 80%, 30% and 10% respectively under soaked condition ??FMWH, 1994). It can be seen that 94 samples do not satisfy the condition for road subgrade, base and sub-base. Hence the CBR from that particular 95 borrow pits are very low. The laboratory tests results for soils from Ekpoma are presented in Table 3. Based on 96 the obtained test results of plasticity, the soil classification was made in accordance to the AASHTO classification system, and it was classified as A-7-5, A-2-6, A-6. From the conventional Atterberg limit tests, liquid limit value 97 ranging from 27.76 to 53.32, plastic limit value of 13.43 to 25.20 and plasticity index value of 11.92 to 34.32. 98 Soils with liquid limit less than 30% are considered to be of low plasticity, those with liquid limit between 30% 99 and 50% exhibit medium plasticity and those with liquid limit greater than 50% exhibit medium plasticity and 100

101 those with liquid limit greater than 50% exhibit high plasticity. The values of California bearing ratio have been

shown in Table 1. It has unsoaked CBR ranges between 10.1 and 33.2, which that of its corresponding soaked samples range between 3.48 and 17.6%. The percentage decreases from soaked CBR to unsoaked CBR. This implies that as water is absorbed into the compacted specimen, the resistance to penetration becomes drastically reduced. It has been recommended by Federal Ministry of Works that the values of CBR for road base, subbase and subgrade should not be less than 80%, 30% and 10% respectively under soaked condition. It can be seen that some of the samples satisfy the condition for road subgrade, but for it to be used for base and subbase materials, it is advisable to improve the soil by stabilization or excavation of the soil.

¹⁰⁹ 8 b) Regression Modelling

For this analysis, geotechnical properties including sieve analysis, liquid limit, plastic limit, plastic index, optimium moisture content and maximum moisture content were taken as independent variables as shown in tables 4 and 5while CBR soaked and unsoaked were taken as the dependent variables. To conduct the multivariate linear regression and solve the regression equation, multivariate statistical software Eview 9.0 was employed. The interphase of the statistical software containing both the dependent and independent variables is presented in tables 6 and 7 representing both the soaked and unsoaked CBR respectively. For ease of data transformation, the selected independent variables were coded as follows

¹¹⁷ 9 Analysis Test Results of soaked and un-soaked CBR for ¹¹⁸ Ebhohimi samples

From the result of Tables 6 and 7, it was observed that the coefficient of determination (R 2) differs for both 119 the soaked and the unsoaked CBR analysis (0.899147 and 0.937723) respectively. The explanation is that the 120 selected independent variables (percent passing 0.075mm sieve size, percent passing 0.425mm sieve size, liquid 121 limit, plastic limit, plastic index, maximum dry density and optimum moisture content) had a better correlation 122 with the unsoaked CBR than the soaked CBR. In addition, the high coefficient of determination as observed 123 revealed the suitability of multivariate linear regression model in explaining the dependence of the independent 124 variables on the regressor. Normally, this would imply a very good fit for the model. Thereafter, multivariate 125 linear regression equation was developed as shown in Figure 2 and 3. The same procedure in Figure 2 applies here. 126 Multiple linear regression equation was developed using the estimated parameters and the substituted coefficients 127 are as shown in Figure 3 which represent the unsoaked CBR. The "Cs" are the soaked CBR coefficient, while 128 X1, X2?Xn are the independent variables (X1 = % 0.075mm sieve, X2 = % 0.425mm sieve, X3 = LL (%), X4 129 = PL (%), X5= PI (%), X6 = MDD (g/cm 3), X7= OMC (%)). The values were substituted and equation 3 130 was derived. Thereafter, a graphical visualization was done, the graphical representation of soaked and unsoaked 131 CBR for Ebhohimi sample, as shown in Figures ?? and 5. ii. 132

10 Statistics of fit based on 95% upper and lower bounds for soaked and unsoaked CBR

The computed statistics of fit based on 95% lower and upper bounds was visualized graphically as presented in Figures ?? and 5

¹³⁷ 11 Comparison of Actual and Predicted CBR Values

From the statistical prediction figures 6 and 7 which shows the actual and predicted soaked and unsoaked CBR 138 based on the multivariate regression approach, it is observed that the actual CBR values and predicted CBR 139 value for both soaked and unsoaked are relatively close, the highest variation is 1.38. To assess the strength of 140 multivariate linear regression analysis in predicting the soaked and unsoaked CBR of the soil based on selected 141 geotechnical parameters, a linear regression of output was done using the actual and predicted soaked and 142 unsoaked CBR as the dependent and independent variables. Result obtained are presented in Figure 8. A plot 143 was made between experimental and predicted values of CBR as shown in Figure 8 It is clear from this figure 144 that most of the predicted CBR values are close to the reported experimental soaked CBR values. As the Actual 145 CBR in soaked and the unsoaked increases, predicted CBR values also increases, indicating linear relationship 146 exists between them. Considering the square of coefficient of correlation (R 2) for both is found to be 0.8991 147 (soaked) and 0.9377 unsoaked, there is evidence that a good correlations exist. 148

¹⁴⁹ 12 c) Ekpoma sample

The input data for Ekpoma analysis is shown in Table ??. 9 and 10, it was observed that the coefficient of determination (R 2) differs for both the soaked and the unsoaked CBR analysis (0.887462 and 0.974403). The selected independent variables (percent passing 0.075mm sieve size, percent passing 0.425mm sieve size, liquid limit, plastic limit, plastic index, maximum dry density and optimum moisture content) had a better correlation with the unsoaked CBR than the soaked CBR. In addition, the high coefficient of determination as observed revealed the suitability of multivariate linear regression model in explaining the dependence of the independent variables on the regressor. From the results, it was observed that 88.7462% and 97.4403% of the variation in

the soaked and unsoaked CBR can be explained by the selected independent variables. Thereafter, multivariate 157 linear regression equation was developed as are shown in The statistical prediction table which shows the actual 158 and predicted soaked and un-soaked CBR based on the multivariate regression approach is presented in figures 13 159 and 14 respectively. The statistical prediction table which shows the actual and predicted soaked and un-soaked 160 CBR based on the multivariate regression approach is presented in Figures 13 and 14 respectively. To assess 161 the strength of multivariate linear regression analysis in predicting the soaked and unsoaked CBR of the soil 162 based on selected geotechnical parameters, a linear regression of output was done using the actual and predicted 163 soaked and unsoaked CBR as the dependent and independent variables. Result obtained is presented in Figure 164 Statistical parameters such as coefficient of multiple determinations (R 2), standard deviation (?), standard 165 error (SE), Adjusted R 2, and mean (μ) of estimated and measured values obtained after multivariate analysis 166 were determined for both soaked and unsoaked CBR for Ebhohimi and, Ekpoma. Comparing the soaked and 167 unsoaked CBR values of these two locations, it was observed in Table 11, that Ekpoma sample has a higher 168 determination coefficient (R 2) of 0.9744 for unsoaked CBR as a function of independent variables (LL, PI, 169 MDD, OMC, 0.075mm and 0.425mm sieve) and Ebhohimi sample has a higher determination coefficient (R 2) 170 of 0.8991 for soaked CBR, which is also as a function of the independent variables. This means that the model 171 has a higher coefficient of determination compared with un-soaked CBR. IV. 172

173 15

174 13 Conclusion

From this study, it was observed that the regression model was able to capture the relation between index properties of soils and the soaked and unsoaked CBRs. At Ebhohimi site, the coefficient of regression with values predicted from the developed regression model and experimentally obtained values were found to be high (Soaked was observed to be 0.89 and the unsoaked is 0.93). Ekpoma (R 2) was observed to be 0.88 for the soaked and 0.97 for the unsoaked.

The results of the analysis indicate that there is a close relationship between experimental CBR values and the predicted CBR values.

However, the results show that more than half of the sample materials do not satisfy the requirement for

both road base and subbase. Some of the materials can only be used as subgrade materials only after thorough compaction by several passes with vibratory roller or excavation and replacement with suitable fill material has

1 2 3 4

185 been carried out.

 $^{^{1}}$ © 2019 Global Journals

 $^{^{2}}$ © 2019 Global JournalsTable 9: Software interphase showing the coefficient estimates of the dependent and independent variables

 $^{^{3}\}text{Regression}$ Modelling of California Bearing Ratio (CBR) Predicted from Index Properties for Lateritic Soils © 2019 Global Journals

⁴Regression Modelling of California Bearing Ratio (CBR) Predicted from Index Properties for Lateritic Soils



Figure 1:

Range: 1 10 - 10 obs Sample: 1 10 - 10 obs Ø c Ø resid Ø soaked_cbr	Filter: * Order: Name	Dependent Variable: SOAKED_CBR Method: Least Squares Date: 10/25/16 Time: 09:51 Sample: 1 10 Included observations: 10				
Y unsoakad_cbr X1 Y2 X1 Y2 Y3 Y4 Y5 Y5 Y5 Y5 Y7 Y7		Variable C X1 X2 X3 X4 X5 X6 X7	Coefficient -31.75696 0.144910 0.204162 0.998236 -0.602393 -1.476291 25.66411 -1.381152	Std. Error 276.8210 0.456320 2.099650 0.588404 0.707574 1.469523 126.3134 3.563582	1-Stalistic -0.114720 0.317563 0.097236 1.696515 -0.851349 -1.004606 0.203178 -0.387574	Prob. 0.9191 0.7809 0.9314 0.2319 0.48429 0.4209 0.8578 0.7357
		R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood F-statistic Prob(F-statistic)	0.899147 0.546161 1.688487 5.701975 -11.38052 2.547259 0.310702	Mean depend S.D. depende Akaike info cri Schwarz criter Hannan-Quin Durbin-Watso	ient var nt var terion rion n criter. m stat	3.100000 2.506379 3.876105 4.118173 3.610557 2.804776

Figure 2: Regression Figure 2 :

Workfile: EBHOHEML - (c:\users\hp\documents\ebhohimi.wf1)	Equation: UNTITLED Work	file: EBHOHIME	E:Ebhohimi\	
View Proc Object Sample Range: 110 - 10 obs Filter: * Sample: 110 - 10 obs Order: Name ØE c - - - Im eq01 - resid -	Dependent Variable: UNSOAH Method: Least Squares Date: 10/25/16 Time: 09:54 Sample: 1 10 Included observations: 10	KED_CBR	imate rorecast stats He	3103
Soaked_cbr Wunsoaked_cbr X x1 X x2 X x3 X4 X5 X5 X5 X5 X7	Variable Corr C 400 X1 0.9 X2 -3.0 X3 0.0 X4 -1.1 X5 0.1 X6 -17 X7 4.0	efficient Sto 97.7347 24 918919 0.4 403428 1.8 623330 0.6 709786 1.3 78.3495 11 014438 3.2	t.statistic 49.5013 1.642214 411285 2.234262 892434 -1.798439 530334 1.175353 637743 -2.818427 324495 0.535892 13.8474 -1.566566 211889 1.249868	Prob. 0.2423 0.1550 0.2139 0.3608 0.1062 0.6457 0.2577 0.3378
← ► Ebhohimi / New Page /	R-squared 0.0 Adjusted R-squared 0.1 S.E. of regression 1.1 Sum squared resid 4.0 Log likelihood -10 F-statistic 4.1 Prob(F-statistic) 0.2	937723 Mear 719755 S.D. 521849 Akali 632047 Schu 0.34145 Hanr 302106 Durb 201525	in dependent var dependent var ike info criterion warz, criterion nan-Quinn criter, bin-Watson stat	3.781000 2.874765 3.668291 3.910359 3.402743 2.843624





Figure 4: Figure 3 :



Figure 5: Figure 4 : Figure 5 :



Figure 6:



Figure 7: Figure 6 :

iew Pr	oc Object	Print Name	Freeze Es	timate	Forecast	Stats	Resids			
obs	Actual	Fitted	Residual		Resid	dual Pl	ot			
1	0.63000	-0.02301	0.65301	1			æ	Т		
2	3.63000	4.26716	-0.63716	1	~	1		- 1		
3	8.18000	7.58368	0.59632	1		\geq	÷	1		
4	2.89000	3.84525	-0.95525	1	\sim	-		1		
5	1.39000	1.44500	-0.05500	1	>	×		1		
6	3.71000	4.62810	-0.91810	1	\sim			- 1		
7	1.91000	1.51727	0.39273	1		≫		1		
8	1.30000	2.07222	-0.77222	1	¢	_		1		
9	6.48000	5.09956	1.38044	1			\sim	ا هج		
10	0.88000	0.56477	0.31523	1		·		1		









Figure 10: Figure 9 : Figure 10 :



Figure 11: Figure 11 :

View Proc Object Save Freeze Details+/- Show Fe Range: 110 10 obs Sample: 110 10 obs	tch Store Delete C	Filter: * Order: Name	Dependent Variable: S Method: Least Squares Date: 10/25/16 Time:	IL Name Freeze OAKED_CBR 11:15	Estimate For	ecast Stats R	esids
wag c Son resid ∭o soaked_cbr ∭o unsoaked_cbr			Sample: 1 10 Included observations:	10			
₩ x1			Variable	Coefficient	Std. Error	t-Statistic	Prob.
N 14 N 13			с	250.7646	109.2050	2.296274	0.148
✓ x4			X1	-0.392245	0.378562	-1.036143	0.409
✓ x5			X2	-0.550976	0.364719	-1.510588	0.270
∑ x0			Х3	-1.532487	1.404143	-1.091403	0.389
			X4	2.405520	2.646556	0.908925	0.459
			X5	1.704945	1.421252	1.199608	0.353
			X6	-106.5613	53.20441	-2.002866	0.183
				-1.800776	1.3/45/5	-1.313098	0.31
			R-squared	0.887462	Mean depend	lent var	9.8990
			Adjusted R-squared	0.493581	S.D. depende	ent var	4.8974
			S.E. of regression	3.485150	Akaike info cri	iterion	5.3254
			Sum squared resid	24.29254	Schwarz criter	rion	5.56753
			Log likelihood	-18.62731	Hannan-Quin	n criter.	5.05991
			Prob(F-statistic)	0.341547	Durbin-Watso	on stat	2.57859

Figure 12: Figure 12 :

Workfile: EKPOMA - (c:\users\hp\documents\ekpoma.wfl) Wiew Proc Object Save Freeze Details+/- Show Fetch Store Delete Genr Sample	Equation: UNTITLED Workfile: EKPOMA::Ekpoma\ View Proc Object Print Name Freeze Estimate Forecast Stats Resids
Range: 110 - 10 obs Filter.* Sample: 10 - 10 obs Order: Name Image: c Order: Image: c Image: Image:	Dependent Variable: UNSOAKED_CBR Method: Least Squares Date: 10/25/16 Time: 11:18 Sample: 1 10 Included observations: 10
winsoaked_cbr	Variable Coefficient Std. Error I-Statistic Prob.
22 12 13 14 15 15 16 17 17	C 274.9037 106.2974 2.566175 0.1226 X1 -1.035559 0.368483 -2.810602 0.1067 X2 -1.363583 0.355008 -3.840991 0.0616 X3 -4.031242 1.366758 -2.949492 0.0883 X4 7.210935 2.576092 2.799176 0.1074 X5 4.594599 1.383411 3.21210 0.0799 X6 -96.96836 51.78786 -1.853106 0.2050 X7 -2.057085 1.337977 -1.537459 0.2840
	R-squared 0.974403 Mean dependent var 14.64100 Adjusted R-squared 0.884815 S.D. dependent var 9.995477 S.E. of regression 3.392359 Akaike info criterion 5.271490 Sum squared resid 23.01620 Schwarz criterion 5.513558 Log likelihood -18.35745 Hannan-Quinn criter. 5.005942 F-statistic 10.87644 Durbin-Watson stat 2.637119 Prob(F-statistic) 0.086759 Schwarz criterion 5.01592
Ekpoma / New Page /	

Figure 13:



Figure 14: Figure 14 :



Figure 15: Figure 15:



Figure 16:



Figure 17:

🔳 Equ	Equation: EQ01 Workfile: EKPOMA::Ekpoma										
View P	roc Object	Print Name	Freeze	Estimate	Forecast	Stats R	esids				
obs	Actual	Fitted	Residual		Resid	lual Plo	t				
1	17.6000	18.4475	-0.84752	1	e			- 1			
2	18.2000	15.9847	2.21532	1				- 1			
3	6.34000	6.85435	-0.51435	1	1	_		- 1			
4	10.9000	12.0383	-1.13827	1	4	+		- 1			
5	8.85000	6.20136	2.64864	1				· 1			
6	3.48000	5.05523	-1.57523	1	~			'			
7	10.1000	10.2143	-0.11429	1		•		- 1			
8	11.1000	9.94660	1.15340	1		>		- 1			
9	4.99000	4.38791	0.60209	1		-		- 1			-
10	7.43000	9.85978	-2.42978	1 4				I			-
						_					-
	4					_					 at



1

Location	Number of Samples	Depth / Chainage
	Collected	
Ebohimi borrow pit	10	1 to 3m
Ekpoma road /BP	10	0.6-3m&43+230-65+100
	Location Ebohimi borrow pit Ekpoma road /BP	LocationNumber of Samples CollectedEbohimi borrow pit10Ekpoma road /BP10

Figure 19: Table 1 :

Regression Modelling of California Bearing Ratio (CBR) Predicted from Index Properties for Lateritic Soils

Figure	20:	Table	2	:
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3

Year 2019												
) Vol-	$\mathrm{S/N}$	Sample	%Passig	% Pas	Liquid	Plastic	Plastic	MDD	OMC	CBR	CBR	AS
ume		Location	$0.075 \mathrm{mr}$	nsing	Limit	Limit	ity	(g/cm)	(%)	Soaked	lUn-	So
XIx X			(no	0.425	(%)	(%)	Index	(3)		(%)	soa	Cl
Issue			200)	mm			(%)				ked (sifi
IV V											%)	ior
ersion												
I							~ . ~ ~				~~ / ~	
Journal	12	EkpomaUjio	36.06	68.35	53.32	19.00	34.32	1.59	18.6	18.2	32.49	A-
of Re-	$3\ 4$	ba RD,1.4m	24.32	65.85	43.81	15.86	27.95	1.62	19.3	6.34	11.4	A-
searches	56	EkpomaUjio ba	36.89	67.99	36.74	15.51	21.23	1.78	8	10.9	12.8	A-
in En-		0.65m Ekpoma	22.15	73.16	27.76	13.43	14.33	1.69	10.8	8.85	12.8	A-
gineer-		Borrow pit 1	23.14	75.49	41.52	13.70	27.55	1.72	13.2	3.48	10.2	A-
ing (0.75m Ekpoma	21.13	76.44	35.76	15.18	20.58	1.76	14.9	17.6	33.2	A-
Ε		Borrow pit 2							13.6			
		0.75m Ekpoma							0			
		Borrow pit 2										
		1.5m Ekpoma										
		Borrow pit 3										
Global	7	1.5m Ekpoma	40.44	74.7	45.80	25.2	14.29	1.59	16.0	8.29	10.1	A-
		50+500										
	8	Ekpoma 43+230	40.74	78.75	33.80	19.74	14.06	1.65	14.0	9.03	11.1	A-
	9	Ekpoma 47+500	36.23	77.43	29.38	17.45	11.92	1.71	13.6	4.99	5.09	A-
	10	Ekpoma 65+100	29.35	73.32	43.81	15.86	27.95	1.62	19.4	7.43	12.67	A-
	© 20	19 Global Journals										

Figure 21: Table 3 :

 $\mathbf{2}$

-	
_	

S/No.	Variable	Variable Definition
	Code	
1	X1	% Passing 0.075mm sieve
2	X2	% Passing 0.425mm sieve
3	X3	Liquid limit (%)
4	X4	Plastic limit (%)
5	X5	Plastic index (%)
6	X6	Maximum dry density (g/cm3)
7	X7	Optimum moisture content (%)

Figure 22: Table 4 :

 $\mathbf{5}$

X1	X2	X3	X4	X5	X6	X7	SOAKED CBR	UNSOAKED CBR
44.1	66.2	57.49	18.57	38.92	1.81	16.92	0.63	0.71
48.1	69.8	56.28	15.5	35.92	1.75	17.30	3.63	5.70
42.1	64.1	54.75	16.4	38.5	1.72	14.30	8.18	9.72
38.2	61.08	46.31	16.97	29.34	1.74	14.30	2.38	2.89
50.66	73.35	56.45	26.11	30.78	1.65	19.76	1.39	1.46
46.4	66.8	54.49	23.71	30.78	1.65	15.20	3.71	4.09
49.1	67.2	58.17	21.23	36.94	1.75	16.81	1.91	6.19
49.9	68.5	53.04	19.46	33.58	1.75	17.34	1.23	1.30
44.5	64.9	46.31	20.19	26.12	1.68	15.54	4.82	6.48
52.16	75.01	54.64	19.78	34.88	1.65	17.40	0.88	1.51

Figure 23: Table 5 :

6

Figure 24: Table 6 :

 $\mathbf{7}$

Figure 25: Table 7 :

8								
X1	X2	X3	X4	X5	X6	X7	SOAKED	UNSOAKED
							CBR	CBR
36.06	68.35	53.32	19	34.32	1.59	18.6	17.6	33.2
24.32	65.85	43.81	15.86	27.95	1.62	19.38	18.2	32.49
36.89	67.99	36.74	15.51	21.23	1.78	10.8	6.34	11.4
22.15	73.16	27.76	13.43	14.33	1.69	13.2	10.9	12.8
23.14	75.49	41.52	13.7	27.55	1.72	14.9	8.85	12.8
21.13	76.44	35.76	15.18	20.58	1.76	13.6	3.48	10.2
40.44	74.7	45.8	25.2	14.29	1.59	16	8.29	10.2
40.74	78.75	33.8	19.74	14.06	1.65	14	9.03	11.1
36.23	77.43	29.38	17.45	11.92	1.71	13.6	4.99	5.09
29.35	73.32	43.81	15.86	27.95	1.62	19.4	7.43	12.67

Figure 26: Table 8 :

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Figure 27: Table 10:

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Location	Model (Y)	R 2	Adjusted R 2	Standard Deviation (?)	Standard Error (SE)	
Ebhohimi CBB-	0.8991 X + 0.3126	0.8991	0.5461	2.5063	1.6884	3.1000
CBRs Ebhohimi	$0.9377 \mathrm{~X} + 0.2355$	0.9377	0.7197	2.8747	1.5218	3.7810
CBRu						
Ekpoma	0.8875 X + 1.114	0.8874	0.4935	4.8974	3.4851	9.8990
CBRs						
Ekpoma	$0.9744 \mathrm{~X} + 0.3748$	0.9744	0.8848	9.9954	3.3923	14.641
CBRu						

Figure 28: Table 11 :

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	Sample No.	BP1	BP2	BP3	BP4	BP5	BP6 BP7		BP8	BP9 BP10	
	Experimental Value of CBRs	0.63	3.63	8.18	2.38	1.39	3.71	1.91	1.30	4.82	0.88
Ebhohimi	Predicted value of CBRs	0.02	4.26	7.58	3.84	1.44	4.62	1.51	2.07	5.09	0.56
	Experimental Value of CBRu	0.71	5.70	9.72	2.89	1.46	4.09	6.19	1.23	6.48	1.51

[Note: () Volume XIx X Issue IV V ersion I]

Figure 29: Table 12 :

- [Kumar ()] 'A Study of Correlation between California Bearing Ratio (CBR) values with other properties of
 Soil'. T Kumar . International Journal of Emerging Technology and Advanced Engineering Certified Journal
- 188 2014. Civil Engineering Department, Nowgong Polytechnic
- [Okovido ()] Civil Engineering History & Phylosophy of Science & Technology, . J O Okovido , MusaM . 2004.
 Benin City: Quality Press Research Studio.
- [Jigar and Patel ()] 'Comparison between soaked and unsoaked CBR'. Sathawara Jigar , K Patel , AK .
 International Journal of Advanced Engineering Research and Studies 2013. 11 p. .
- ¹⁹³ [Forkenbrock and Weisbrod ()] 'Correlation of CBR Values with Soil Index Guide for Mechanistic and Empirical
- Design for New and Rehabilitated 13. Pavement Structures". Final Document'. D Forkenbrock , Glen E
 Weisbrod , GE . National Cooperative Highway Research Program 2001. 2001. (Properties. West University
 Avenue)
- ¹⁹⁷ [Carter and Bentley ()] Correlations of soil properties, M Carter, S P Bentley . 1991. London: Pentech Press. ¹⁹⁸ p. 130.
- [General Specification for Roads and Bridges ()] General Specification for Roads and Bridges, 1994. (Federal
 Ministry of Works and Housing)
- 201 [Her Majesty's Stationery Office] Her Majesty's Stationery Office, London. p. 673.
- [Methods of test for soils for civil engineering purposes ()] Methods of test for soils for civil engineering purposes,
 BS 1377-2) and Part 4 (BS 1377-4): BS 1377. 1990. BS, London. 2. BSI (British Standard Institution
- 204 [References Références Referencias] References Références Referencias,
- [Bello ()] 'Regression analysis between properties of subgrade lateritic soil'. A A Bello . Leonardo Journal of
 sciences 2012. 21 p. .
- [Orie and Nweni ()] 'Severity indices of variables causing road collapse in Nigeria'. O U Orie , E L Nweni .
 Journal of Civil and Environmental System Engineering 2015. 13 p. 156. University of Benin
- [Arora ()] Soil Mechanics and Foundation Engineering, Re-print Standard Publishers Distributer, K R Arora.
 2004. 2004. Nai Sarak, Delhi.
- 211 [Standard Practice for classification of Soils for Engineering Purposes (Unified Soil Classification System) American Society for T 212 'Standard Practice for classification of Soils for Engineering Purposes (Unified Soil Classification System)'.
- 213 American Society for Testing and Materials 2000. 04 p. 8. (Annual Book of ASTM Standards)
- [Croney ()] The design and performance of road pavements, D Croney . 1977. Department of the Environment.
 Department of Transport: Transport and Road Research Laboratory
- [Alayaki ()] 'Water Absorption Properties of Lateritic Soil in Road Pavement. A Case Study of Ife-Ilesha
 Highway, South-Western Nigeria', F M Alayaki . International Journal of Emerging Technology and Advanced
- ²¹⁸ Engineering 2012. p. 52.