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## 1.Introduction Regression Analysis

### 1.1 General Information

Regression analysis, two or more variables with cause-effect relationship between them to determine the relationship between and estimation using that relationship or to make predictions. Numerical correlation between a criterion variable and one or more forecast variables It is the statistical analysis used in converting into halo. Regression analysis mainly variables It aims to determine the nature of the relationship between. A variable as a prediction variable simple regression, if used, two or more variables as prediction variables If used, it is possible to talk about multiple regression analysis. The goal is every guess variable's contribution to the total change in the criterion variable and hence the criterion based on the value of the linear combination of prediction variables value is estimated. Regression with one dependent variable and more than one independent variableRegression with one dependent variable and more than one independent variable models are called multivariate regression analysis. The change of independent variables simultaneously (simultaneously) in the dependent variable.I try to explain the univariate regression analysis in terms of computation and interpretation.Interpretation of multivariate regression analysis to univariate regression analysisHowever, there are some differences. For example, in univariate regression analysis multiple regression coefficient is expressed as R (multiple R). The coefficient R is multiple (simultaneously) handled simultaneously with the change in a dependent variable.shows the degree of the relationship between the change in the independent variable. In other words, the change in a group of independent variables that are considered together with the dependent variableIt is an indication of the relationship (correlation). Multivariate regression analysis in many branches of social sciences Behavioral movements in branches of science such as marketing, sociology and psychology.determining the factors affecting the economic variables of the time series type in the economy. It finds use in its detection and projections (forecasting) for the future.

## 1.1.1 Multiple Regression Analysis Models

- Enter method: Entering arguments as a block in one step method in which it is evaluated.
- Forward selection: Highest with dependent variable the positive or negative correlated argument is selected first. The hypothesis that the coefficient of the entered variable is 0 is examined with the F test. The F value is compared to the predicted F values of SPSS. SPSS has two F criteria. The F value is compared with a minimum F value that you determine. (F-toenter, FIN). The normal setting is 3.84. "Use F Value" in the window passed through the "options" key. It is part. Setting the probability associated with the F statistic (Porbability of F-toenter, PIN). The normal setting for this is 0.05. If the F value obtained is less than or equal to these criteria, that argument the regression is taken into consideration and the selection continues forward, otherwise the process is there. It is stopped.

- Backward Elimination Method: Forward selection on the contrary, here all arguments are selected first, and then in order to certain criteria. Screening is done according to SPSS; it sets two criteria for screening. The least squares F value (F-toremove, FOUT) to keep the variable in the formula. Normal setting is 2.71. The greatest probability of F (Probabability of F-to-remove, POUT). Setting is 0.10. First, the variable with the smallest partial correlation coefficient is examined. A variable with a large value is eliminated from the values.
- Stepwise Selection Method: First argument if this meets the FIN or PIN requirements in the right choice, the second variable is selected, or the process ends there. The variable with partial correlation is taken. Selections from high correlation to low. If the independent variables meet the criteria, regression analysis is started. Selecting Step by Step after the variable is selected differs from choosing forward. The variable matches the FOUT or POUT criteria as in the backward elimination so it is checked both step by step and forward selection. Reverse sieving operations are done.

## 1.2 Analysis problem and aim

In this research, I decided to do it because of the incident we have experienced in recent months. In recent months, I think that there has been a decrease in the level of people happiness due to earthquakes and viruses. So I decided to investigate the happiness level of the first hundred countries of 2019, and one of my reasons for this research was to make people understand that it is necessary to work so that they do not become unhappy every year. In this study, we take the happiness score as a dependent variable and look at the factors affecting the happiness score. I think other factors that I think affect the happiness score are GDP per capita, Social support, Freedom to make life choices, and Perceptions of corruption, so independent variables are GDP per capita, Social support, Freedom to make life choices, and Perceptions of corruption.

## 1.3 Define Data Set

This is my data set definetion. In this all data are numeric but country is string, and country didnot effect regression analysis. Also score is dependent variable, and other variables are independent. Also in this research used to enter method.

Name	Туре	Width	Decimals	Label	Values
Country orr	String	22	0	Country or reg	None
Score	Numeric	18	15		None
GDPperca	Numeric	19	15	GDP per capita	None
Socialsupport	Numeric	19	15	Social support	None
Freedomto	Numeric	21	15	Freedom to m	None
Perception	Numeric	21	15	Perceptions of	None

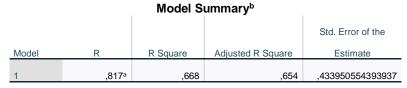
#### **Descriptive Statistics**

	Mean	Std. Deviation	N
Score	6,072569999999998	,737496012726767	100
GDP per capita	1,111710000000000	,274388145597517	100
Social support	1,362510000000000	,172099307424991	100
Freedom to make life choices	,435210000000000	,122162443186327	100
Perceptions of corruption	,118720000000000	,107420732344244	100

This table figure out my date set variables mean, standard deviation, and values of variables.

## 2. Regression Analysis

## 2.1 Model Summary



a. Predictors: (Constant), Perceptions of corruption, Social support, Freedom to make life

choices, GDP per capita

b. Dependent Variable: Score

R-squared is a statistical measure of how close the data are to the fitted regression line. It is also known as the coefficient of determination, or the coefficient of multiple determination for multiple regression. The definition of R-squared is fairly straight-forward; it is the percentage of the response variable variation that is explained by a linear model. Or:

R-squared = Explained variation / Total variation

R-squared is always between 0 and 100%:

- 0% indicates that the model explains none of the variability of the response data around its mean.
- 100% indicates that the model explains all the variability of the response data around its mean.

In general, the higher the R-squared, the better the model fits your data. In myresearch R squared is %66 but we use adjusted R square, because we use multiple regression analysis, so this R value was %65. Also this value sad that Taken as a set the predictors GDP per capita, Social support, Freedom to make life choices, and Perceptions of corruption for %65 of the variance in score.

#### 2.2 Anova Table

ANOVA <sup>a</sup>								
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	35,956	4	8,989	47,735	,000 <sup>b</sup>		
	Residual	17,890	95	,188				
	Total	53,846	99					

a. Dependent Variable: Score

b. Predictors: (Constant), Perceptions of corruption, Social support, Freedom to make life choices, GDP per capita

ANOVA is used to compare differences of means among more than 2 groups. It does this by looking at variation in the data and where that variation is found (hence its name). Specifically, ANOVA compares the amount of variation between groups with the amount of variation within groups. It can be used for both observational and experimental studies. Here is my Anova table. Anova table is a test of whether that R square is greater than 0. My P value is less than 0.05. So that means that the test is significant, regression is significant. Also that means my predictors are able to account for a significant amount of variance in score. My regression model is significant. So i can predict score significantly. Also The difference between the observed value of the dependent variable (y) and the predicted value ( $\hat{y}$ ) is called the **residual** (e). Residual = Observed value - Predicted value  $e = y - \hat{y}$  (in general). Also in regression part df 4 is related to independent variables. Residual df is the total number of observations (rows) of the dataset subtracted by the number of variables being estimated. **Regression SS** is the total variation in the dependent variable that is explained by the regression model. It is the sum of the square of the difference between the predicted value and mean of the value of all the data points. **Residual** SS — is the total variation in the dependent variable that is left unexplained by the regression model. It is also called the **Error Sum of Squares** and is the sum of the square of the difference between the actual and predicted values of all the data points. Mean Squared Errors (MS) are the mean of the sum of squares or the sum of squares divided by the degrees of freedom for both, regression and residuals. Regression MS =  $\sum$  ( $\hat{y}$  —  $\bar{y}$ )<sup>2</sup>/Reg. Df Residual MS =  $\sum$  (y —  $\hat{y}$ )<sup>2</sup>/Res. Df.**F** — is used to test the hypothesis that the slope of the independent variable is zero. Mathematically, it can also be calculated as F = Regression MS / Residual MS. This is otherwise calculated by comparing the F-statistic to an F distribution with regression df in numerator degrees and residual df in denominator degrees. Significance F — is nothing but the p-value for the null hypothesis that the coefficient of the independent variable is zero and as with any pvalue, a low p-value indicates that a significant relationship exists between dependent and independent variables.

This overall regression model is significant.

GDP per capita

Social support

Freedom to make life choices

Perceptions of corruption

$$F(4,95) = 48,p < 0.01, R Square = .668$$

### 2.3 Correlations

					Freedom to make	Perceptions of
		Score	GDP per capita	Social support	life choices	corruption
Pearson Correlation	Score	1,000	,646	,620	,530	,638
	GDP per capita	,646	1,000	,579	,201	,496
	Social support	,620	,579	1,000	,324	,364
	Freedom to make life choices	,530	,201	,324	1,000	,484
	Perceptions of corruption	,638	,496	,364	,484	1,000
Sig. (1-tailed)	Score		,000	,000	,000	,000
	GDP per capita	,000		,000	,023	,000
	Social support	,000	,000		,001	,000
	Freedom to make life choices	,000	,023	,001		,000
	Perceptions of corruption	,000	,000	,000	,000	
N	Score	100	100	100	100	100

Correlations

This table show that correlation between each variable. The important correlation is score between other variables, because score is dependent variable in this research. Also statistics expert sad that your correlation values must not be over .800. I look the my values. This are not over than .800.

		Coe	efficients				
			Standardized				
	Unstandardize	ed Coefficients	Coefficients			95,0% Confider	ce Interval for B
Model	В	Std. Error	Beta	t	Sig.	Lower Bound	Upper Bound

1	_(Constant)	2,732	,369		7,401	,000	1,999	3,464
	GDP per capita	,843	,213	,314	3,954	,000	,420	1,266
	Social support	1,113	,322	,260	3,459	,001	,474	1,752
	Freedom to make life	1,544	,421	,256	3,671	.000	.709	2,379
	choices	·	,	·	,	,	,	,
	Perceptions of corruption	1,812	,524	,264	3,456	,001	,771	2,852

a. Dependent Variable: Score

### 2.4 Coefficients

Regression coefficients represent the mean change in the response variable for one unit of change in the predictor variable while holding other predictors in the model constant. This <u>statistical control</u> that regression provides is important because it isolates the role of one variable from all of the others in the model. The p-value for each term tests the null hypothesis that the coefficient is equal to zero (no effect). A low p-value (< 0.05) indicates that you can reject the null hypothesis. In other words, a predictor that has a low p-value is likely to be a meaningful addition to your model because changes in the predictor's value are related to changes in the response variable,so I evaluated each of these test for alpha of .05. Also we look to significant level. If their significant level is less than 0.05, we sad that it is significant. Standard Error — provides the estimated standard deviation of the distribution of coefficients. It is the amount by which the coefficient varies across different cases. A coefficient much greater than its standard error implies a probability that the coefficient is not 0.

**t-Stat** — is the t-statistic or t-value of the test and its value is equal to the coefficient divided by the standard error.

#### t-Stat = Coefficients/Standard Error

Again, the larger the coefficient with respect to the standard error, the larger the t-Stat is and higher the probability that the coefficient is away from 0.

**p-value** — The t-statistic is compared with the t distribution to determine the p-value. We usually only consider the p-value of the independent variable which provides the likelihood of obtaining a sample as close to the one used to derive the regression equation and verify if the slope of the regression line is actually zero or the coefficient is close to the coefficient obtained.

A p-value below 0.05 indicates 95% confidence that the slope of the regression line is not zero and hence there is a significant linear relationship between the dependent and independent variables. A p-value greater than 0.05 indicates that the slope of the regression line may be zero and that there is not sufficient evidence at the 95% confidence level that a significant linear relationship exists between the dependent and independent variables.

Lower and Upper 95% — Since we mostly use a sample of data to estimate the regression line and its coefficients, they are mostly an approximation of the true coefficients and in turn

the true regression line. The lower and upper 95% boundaries give the 95th confidence interval of lower and upper bounds for each coefficient.

Score Significant p = .000

GDP per capita Significant p=.000

Social Support Significant p=.001

Freedom to make life Choices Significant p=.000

Perceptions of Corruption Significant p=.001

### 3. CALCULATION OF FORECAST

Recall that predictions of yy can be obtained using $^{t}=^{\beta}0+^{\beta}1x1,t+^{\beta}2x2,t+\cdots+^{\beta}kxk,t,yt^{=\beta}0+^{\beta}1x1,t+^{\beta}2x2,t+\cdots+^{\beta}kxk,t,which$ comprises the estimated coefficients and ignores the error in the regression equation. Plugging in the values of the variables x1,t,...,xk,tx1,t,...,xk,t for t=1,...,Tt=1,...,T returned the fitted (training-sample) values of yy. What we are interested in here, however, is forecasting future values of yy.

First of all let us calculate estimation of row of my dataset;

According to formula and my coefficients table

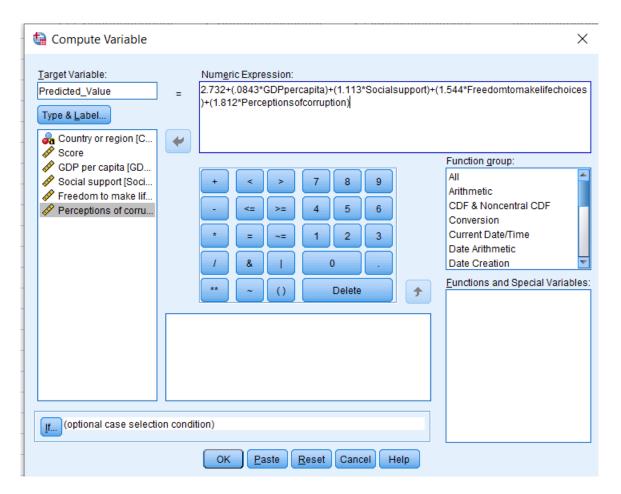
	Coefficients <sup>a</sup>							
				Standardized				
		Unstandardize	d Coefficients	Coefficients			95,0% Confiden	ce Interval for B
Model		В	Std. Error	Beta	t	Sig.	Lower Bound	Upper Bound
1	(Constant)	2,732	,369		7,401	,000	1,999	3,464
	GDP per capita	,843	,213	,314	3,954	,000	,420	1,266
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	Freedom to make life	1,544	,421	,256	3,671	,000	,709	2,379
	choices							
	Perceptions of corruption	1,812	,524	,264	3,456	,001	,771	2,852

a. Dependent Variable: Score

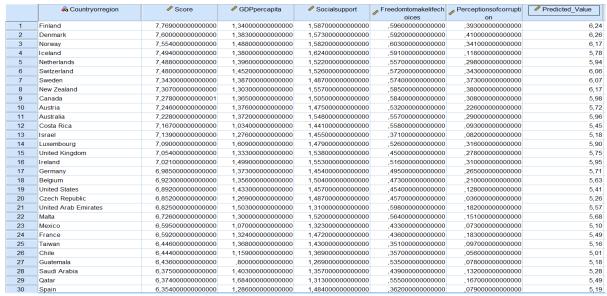
B1=2.732 B3=1.113 B5=1.812

B2=.0843 B4=1.544

Then i calculated my predictions via SPSS. In order to make that calculation i clicked "Transform" from top menu then i clicked "Compute Variable" option.



After that all my predictions are calculated. As you can see down there my predictions are under the new variable which is "Predicted Value".



Also, we see that the predicted values are roughly close to the scores. This shows that our prediction is almost correct.

# 3.1 FORECASTING'S CONFIDENCE INTERVAL

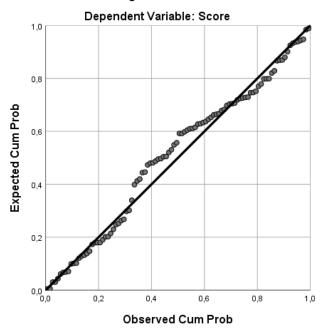
## **Descriptives**

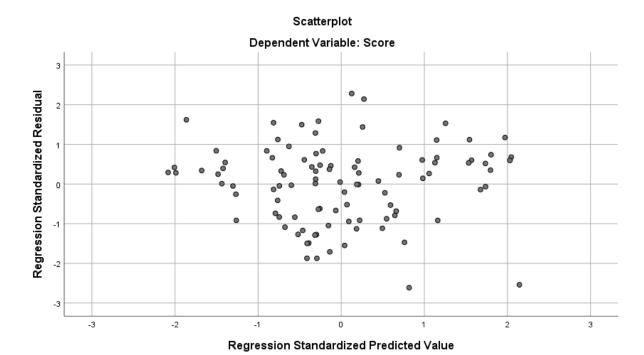
			Statistic	Std. Error
Predicted_Value	Mean		5,2293	,04562
	95% Confidence Interval for Mean	Lower Bound	5,1388	
		Upper Bound	5,3198	
	5% Trimmed Mean	5,2218		
	Median	5,1929		
	Variance	,208		
	Std. Deviation	,45620		
	Minimum	4,28		
	Maximum	6,26		
	Range	1,97		
	Interquartile Range	,52		
	Skewness	Skewness		
	Kurtosis		-,153	,478

Then i got the results. Here %95 Confidence Interval for my prediction values. Lower Bound is 5.1388. Upper bounds is 5.3198. %95 of all values from my predicted values are falling between 5.1388 and 5.3198.

# **4.CHECK THE ASSUMPTION**

Normal P-P Plot of Regression Standardized Residual





This show that heteroscedasticity, because Heteroscedasticity means unequal scatter. In <u>regression analysis</u>, we talk about heteroscedasticity in the context of the <u>residuals</u> or error term. Specifically, heteroscedasticity is a systematic change in the spread of the residuals over the range of measured values. Heteroscedasticity is a problem because <u>ordinary least squares</u> (<u>OLS</u>) <u>regression</u> assumes that all residuals are drawn from a <u>population</u> that has a constant variance (homoscedasticity).