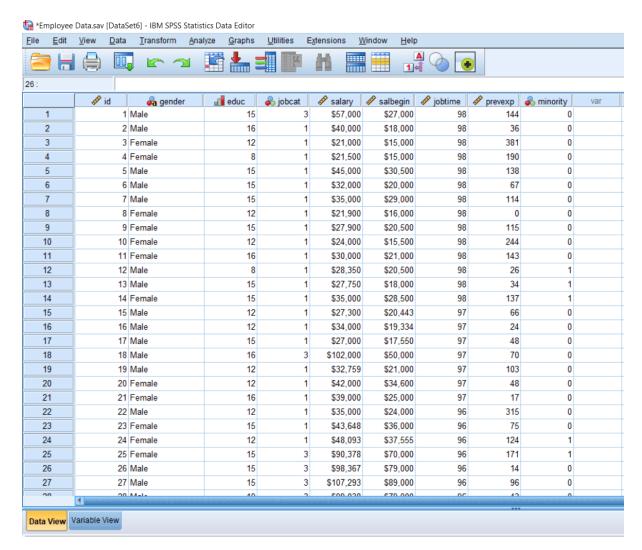
One-Way ANOVA

One-Way ANOVA in SPSS

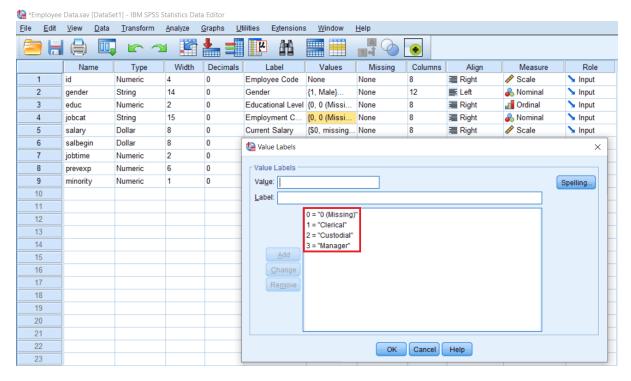
- In this section, we will learn how to conduct One-Way ANOVA in SPSS. The theory of One-Way ANOVA was given by British Mathematician or statistician Ronald Fisher. He gives One-Way ANOVA to identify the significant differences between more than two groups. Whenever we have more than two groups, we cannot apply a t-test. So, in that case, we have to go to One-Way ANOVA. Generally, more than two groups mean more than two levels of one independent variable.
- For example, suppose there is a researcher who wants to determine the effect of three different kinds of therapies or medicines on the patient's condition, such as depression. Suppose he treats the patient for depression and has three or more than three therapies. In that case, we want to determine whether these three different therapies lead to significantly different results. We will check whether one therapy leads to significantly more improvement in the patient's condition as compared to other therapy. In this case, we have just two variables. The first variable is a therapy that is our independent variable, and the second variable is depression, that is our dependent
- Therapy has three levels, which are Psychodynamic therapy, behavioral therapy, and cognitive So we have three levels of a single variable. In that case, we want to determine whether these three different levels have a significantly different influence on the dependent variable.

Calculating One-Way ANOVA in SPSS

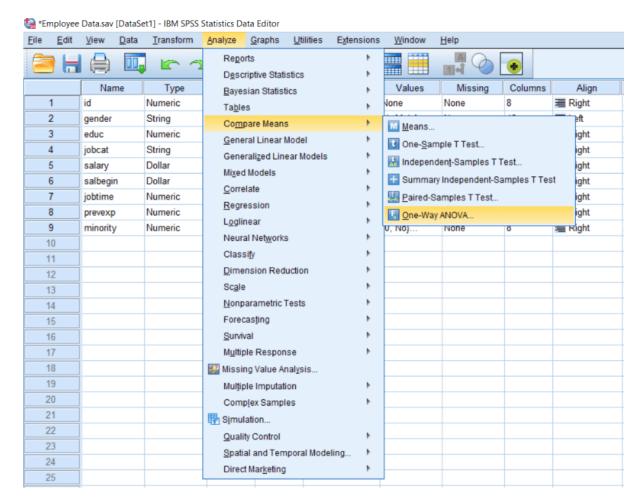
To calculate One-Way ANOVA, we will take the **Employee data set**, which is given below:



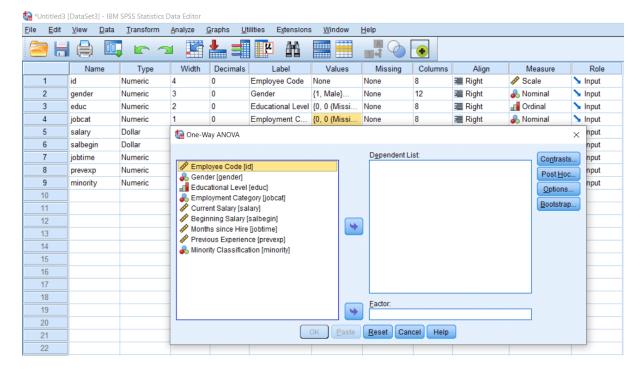
In this data set, we have a variable called **jobcat**. Now, if we look at the **jobcat variable**, we find that jobcat has **three levels**. If we click on the Values option, there are people from three different job categories, i.e., **Clerical** level, **Custodial** level, and **Managerial** level. **0** stands for the **missing value**. That is a good way of defining missing value. O has been defined as a missing value, and they have kept in the missing value option. It means if we calculate any **stat**, for example, we want to do **One-Way ANOVA**, 0 will not be counted as a variable.



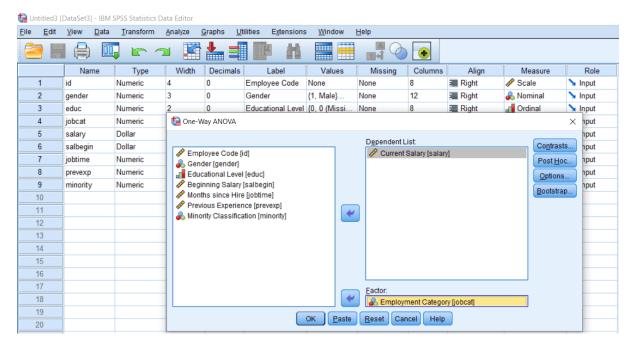
We want to find out that people belonging to different job categories will be drawing significantly different amounts of salary. To be more precise, we want to make a test. In which we can say that Managers are drawing significantly more amount to salary as compare to custodial or clerical people. So we hypothesize that when we move from the clerical level to the managerial level, there is an increase in salary. We can test it by using One-Way ANOVA. So we have two variables here. One is an independent variable that is jobcat, which has three levels and one dependent variable, i.e., current salary. Now once all variables are defined, we will move to the calculation of One-Way ANOVA. To calculate One-Way ANOVA, we will go Analyze menu, and Compare Means and then One-Way ANOVA.



The symbol of **One-Way ANOVA** is **F a**. So **F** stands for the **Ronald fisher**, the person who gives this test. When we click on **One-Way ANOVA**, we will see a dialog box like this:



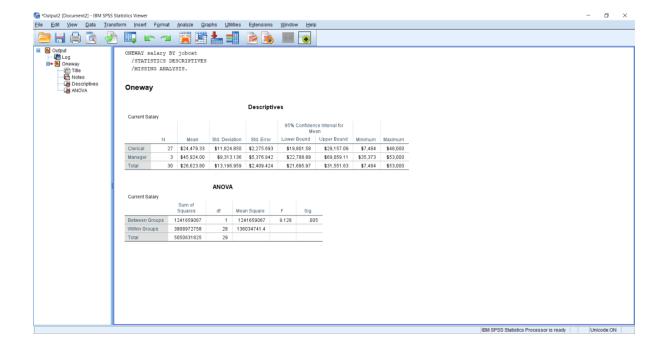
All the variables are populated on the **left-hand side**. So we have to select our **dependent** variable in the **Dependent List** and **independent** variable in **Factor**. So our **dependent** variable is the **Current salary**, and the **independent** variable is **Job category**, so we will select it.



By going to the **Options tab**, we can look at the **Descriptive statistics** like this:

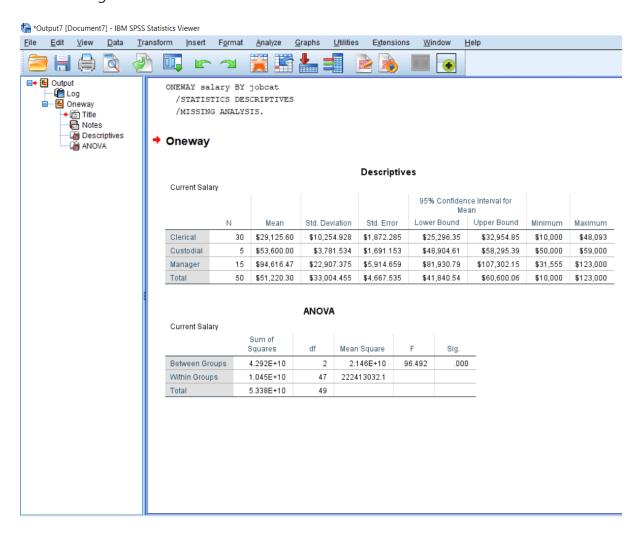


Now we are ready to test the significant differences between three **salary groups**, three **job categories**. So click on, **Ok**. The following is the **output** for **One-Way ANOVA**.

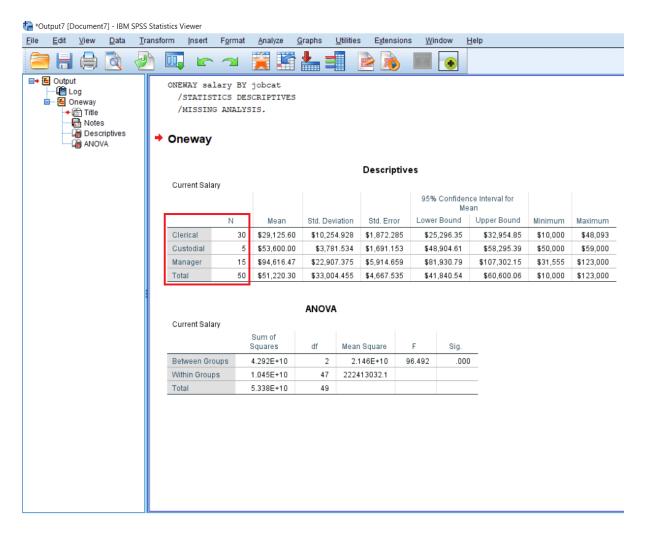


Output of One-Way ANOVA

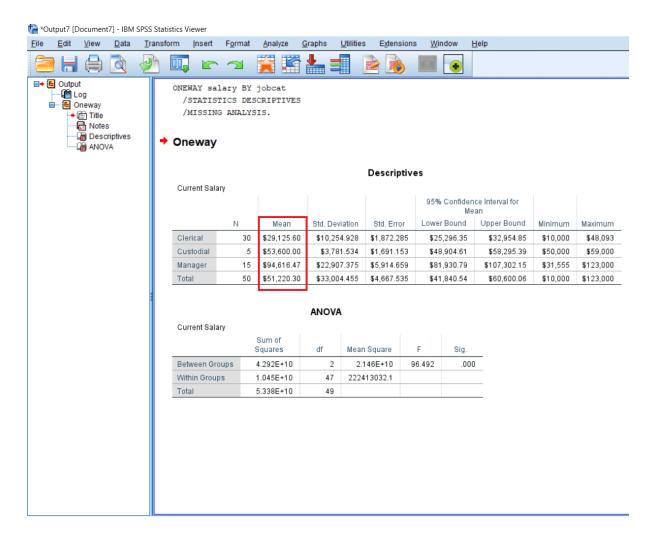
In this section, we will learn the output of **One-Way ANOVA**. The output of **One-Way ANOVA** is given below:



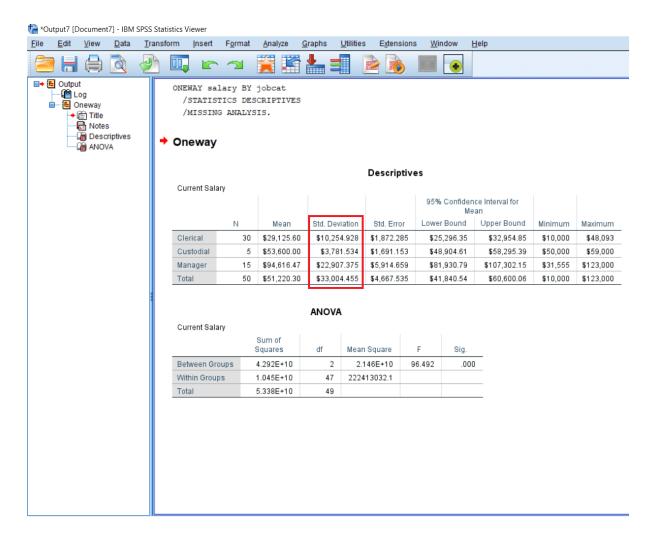
In the above output, we can see only three **job categories** are appearing, i.e., **Clerical**, **Custodial**, and **Manager**. The missing value **0** is not available in the output. **30** people belong to job **category 1**, i.e., **Clerical**, **5** belonging to job **category 2**, i.e., **Custodial**, and **15** belonging to job **category 3**, i.e., **Managers**. We are having a slightly lesser number of people in job category 2.



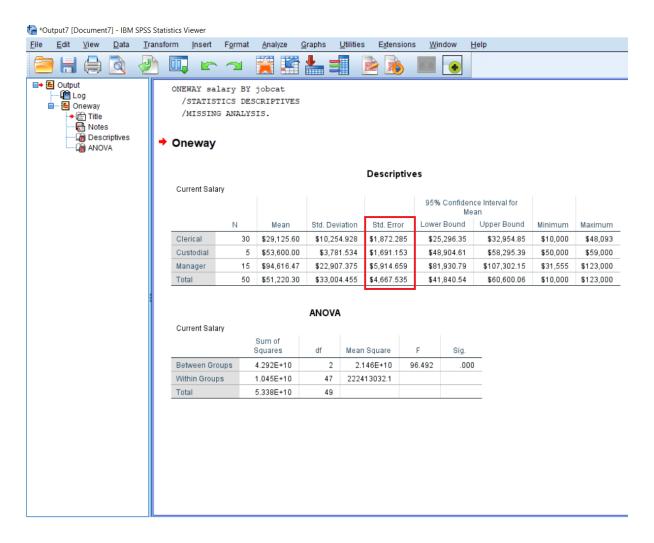
will it any recursion for see has our analysis. Average salaries are 29,125 dollars, 53,600 dollars, and 94,616 dollars. If we these **Means scores**, we will see the impression that there are differences between people's salaries from three different job categories. But if we move from Clerks to Custodial employees, the amount of difference is not huge. But if we move from the Custodial employees to the Manager, there is a huge significant difference, more than double the salary difference. So, we are expecting a significant difference between the groups. So that's how we can guess about the significant differences just by looking at the **Mean scores**.



The **Standard deviation** in the case of **Managers** is quite **high** compared to the rest of two groups. So, managers are drawing on an average more salary, but there is a huge variation in the managerial category compared to other groups.



The **Standard error** is shown in the following image. **Standard error** refers to the standard deviation of the sampling distribution of mean. So, it's an indication of the amount of error in measurement. So, the smaller it is better for us.



There is **95% confidence information**. So, we can see none of the confidence information is having 0. So, we don't have a positive or negative value on either side, and these are the minimum and maximum amount of salary drawn by different groups. So, in the case of **Manager**, the difference is very high, **31 thousand** to **123 thousand** that's the descriptive scenario of output.

