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SPSS TUTORIAL & EXERCISE BOOK
FOR BUSINESS STATISTICS

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INTRODUCTION

This exercise book was written for the students of the University of Miskolc within the framework of Business Statistics and Quantitative Statistical Methods. Some parts of the exercises are translated from the Hungarian book of Domán – Szilágyi – Varga: *Statisztikai elemzések alapjai II*, which are supplemented by SPSS exercises on the basis of SPSS 16.0 and 19.0 Tutorial. This book is a tutorial, which includes theoretical background just to understand the examples included.

ACKNOWLEDGEMENTS

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I. SPSS TUTORIAL

1. INTRODUCTION TO SPSS¹

Open the *Csernely_data.sav* file!

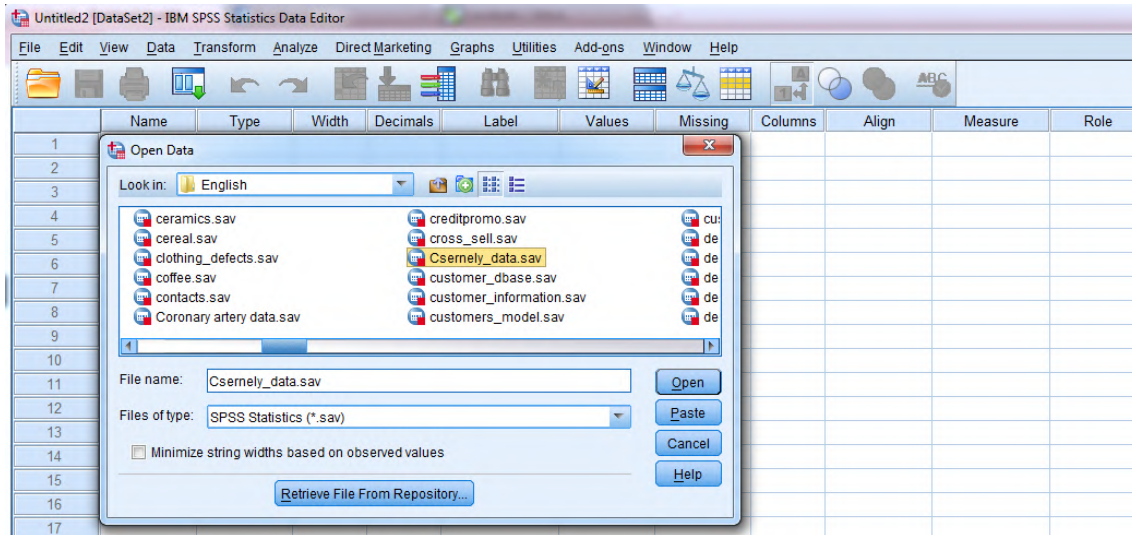


Figure 1 – Opening an SPSS Data File

The Data Editor provides 2 views of data: the *Data View* and *Variable View*. Information can be edited or deleted in both views.

Data View: this view displays the actual data values or value labels.

Variable View:

- **Name:** it is a unique name of each variable (the names should be different). The names cannot contain space or other illegal characters and the first character must be a letter.
- **Type:** it specifies the type of data for each variable. The original setting is the most frequently used type, the numeric type, which refers to variable, whose values are numbers. However, we can change to Comma, Dot, Scientific Notation, Date, Dollar, Custom Currency or String Variables.

¹ According to SPSS 16.0 Tutorial

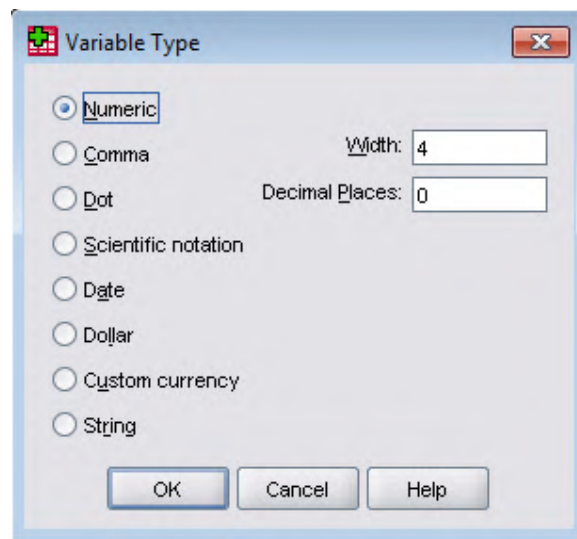


Figure 2 – Variable Type Dialog Box

- Width: the field width.
- Decimals: number of decimals in case of Numeric type.
- Label: descriptive name of a variable (up to 256 characters). It can contain space or other characters, which we could not use in Names.
- Values: we can assign descriptive value labels for each value of a variable, thus the numeric codes represent non-numeric categories.

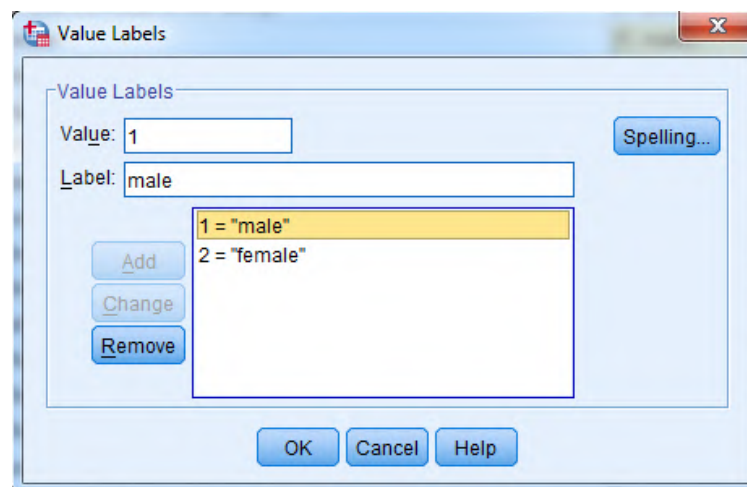


Figure 3 – Value Labels Dialog Box

- Missing: if we do not have data, because e.g. a respondent refused to answer. User-missing values are flagged for special treatment and are excluded from most calculations.

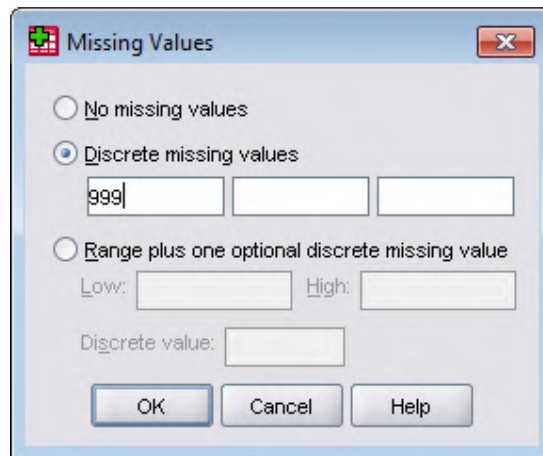


Figure 4 – Missing Values Dialog Box

- Column: number of characters for the column width.
- Align: alignment controls the display of data. It can be right, left or center.
- Measures: it is the scales of measurement, which can be nominal, ordinal, interval or ratio scale. In the SPSS you will find the nominal, ordinal and ratio measures.
 - Nominal scale: Numbers are labels or groups or classes. Simple codes assigned to objects as labels. We use nominal scale for qualitative data, e.g. professional classification, geographic classification. E.g. blonde: 1, brown: 2, red: 3, black: 4. A person with red hair does not possess more ‘hairiness’ than a person with blonde hair.
 - Ordinal scale: Data elements may be ordered according to their relative size or quality, the numbers assigned to objects or events represent the rank order (1st, 2nd, 3rd etc.) E.g. top lists of companies.
 - Interval scale: There is a meaning of distances between any two observations. The "zero point" is arbitrary. Negative values can be used. Ratios between numbers on the scale are not meaningful, so operations such as multiplication and division cannot be carried out directly. E.g. temperature with the Celsius scale.
 - Ratio scale (Scale): This is the strongest scale of measurement. Distances between observations and also the ratios of distances have a meaning. It contains a meaningful zero. E.g. mass, length.

2. TRANSFORM / SELECT DATA

Example 1 How old are the respondents? Create a new variable as *age*!

We can create new variables by transforming another one. We have data about the date of birth of employees. If we subtract the year of birth from the current year, we will get their age. In order to determine the age, choose *Transform / Compute Variable...* from the menus.

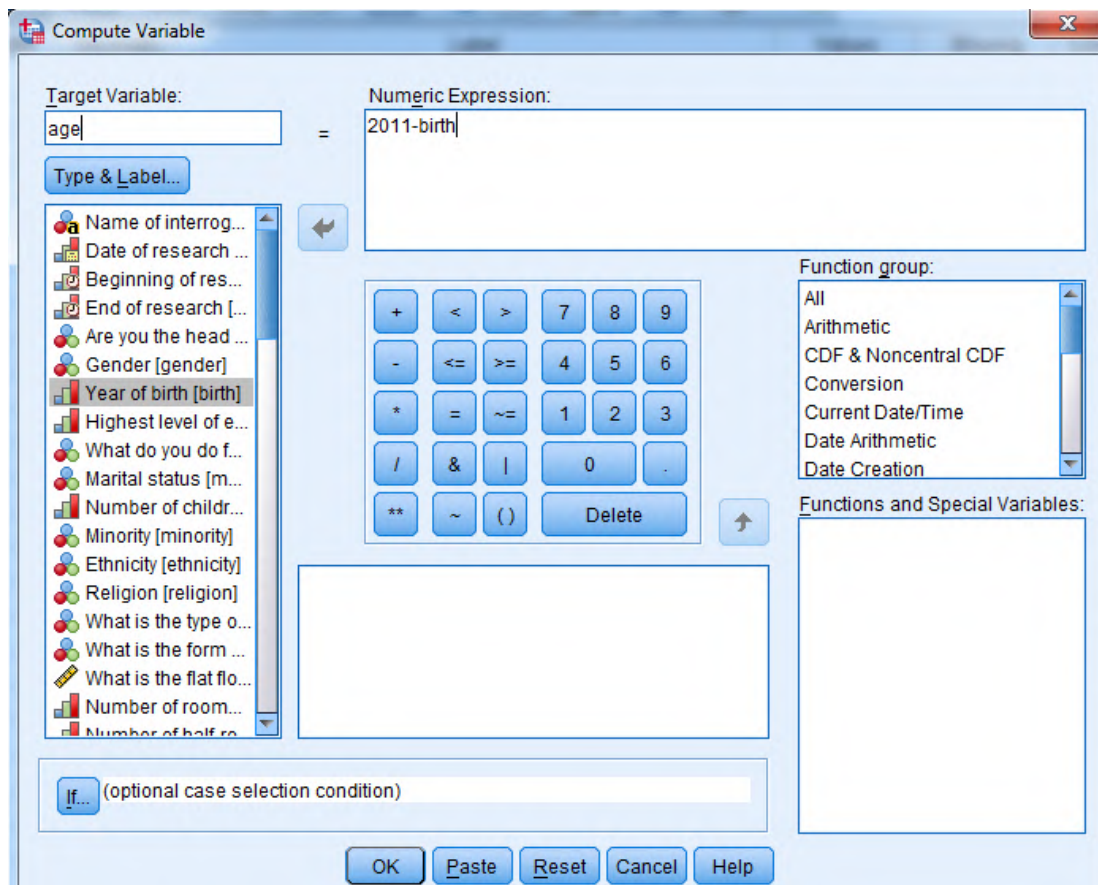


Figure 5 – Compute Variable Dialog Box

Type the name of target variable, say age. To build an expression, type components directly in the Expression field. If the date of birth is given as Date (mm/dd/yyyy), we need just the year part of this. Thus we should extract date.

When we are ready with the expression, press OK, then the new variable will be ready.

Example 2 What is the proportion of single people?

From the menus choose: *Analyze / Descriptive Statistics / Frequencies...* Select the variable, which relative frequency should be calculated (*Marital status*), and then press OK.

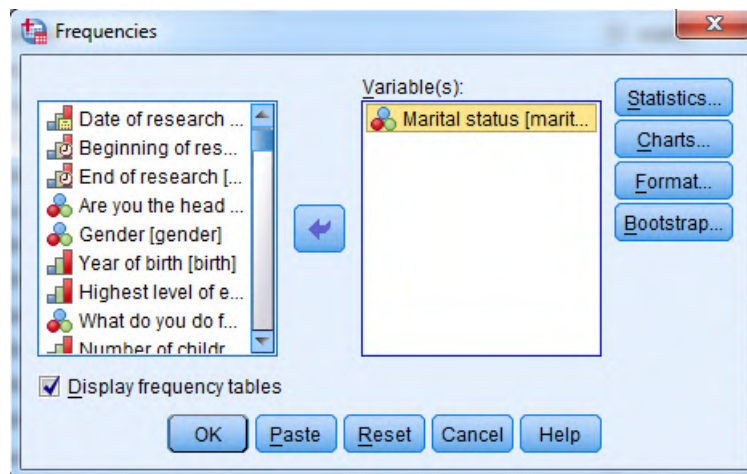


Figure 6 – Frequencies Dialog Box

Find the results in the Output View.

		Marital status			
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	single	12	5.4	5.5	5.5
	married	102	45.9	46.6	52.1
	divorced	27	12.2	12.3	64.4
	partner (but not married)	21	9.5	9.6	74.0
	widow	57	25.7	26.0	100.0
	Total	219	98.6	100.0	
Missing	System	3	1.4		
Total		222	100.0		

Therefore, 5.4% of people are single in Csernely. (5.5% of respondents are single.)

Example 3 What is the proportion of men within pensioners?

Now, the statistical population is not the respondents, but just pensioners. First we should select the subset of cases (*pensioner*) with *Data / Select cases...*

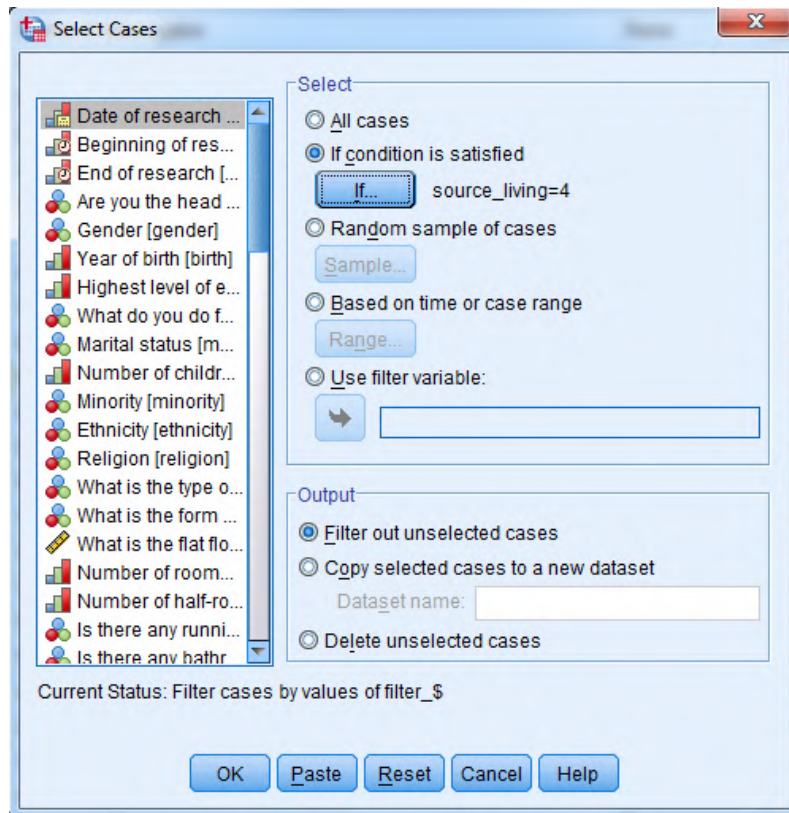


Figure 7 – Select Cases Dialog Box

We use a conditional expression to select men: Gender = 1 (because 1 is the code of men).

	jobcat
1	unemployment benefit
2	.
3	pensioner
4	pensioner

Figure 8 – Selected Subset of Cases

Then choose *Analyze / Descriptive Statistics / Frequencies...* The relative frequencies of gender are the question, so Gender should be added to Variables. The following are the results found in the Output view:

		Gender			
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	Male	62	50.4	50.4	50.4
	Female	61	49.6	49.6	100
Total		123	100.0	100.0	

50.4% of pensioners are men.

3. GRAPHS

Example 1

Create a bar chart about the proportion of respondents grouped by *gender*! Embellish the graph! Attach the value of proportions to the chart!

For creating a bar chart, choose *Graphs / Legacy Dialogs / Bar Charts / Simple* from menus.

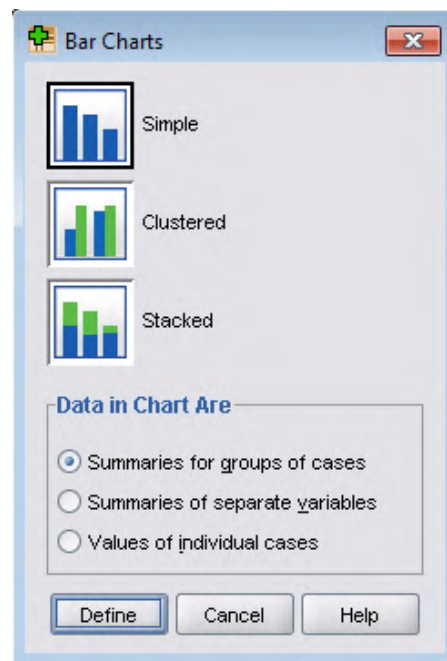


Figure 9 – Bar Charts Dialog Box

Then select a variable for the category axis (*gender*). The question was about the proportion of respondents, therefore bars should represent % of cases.

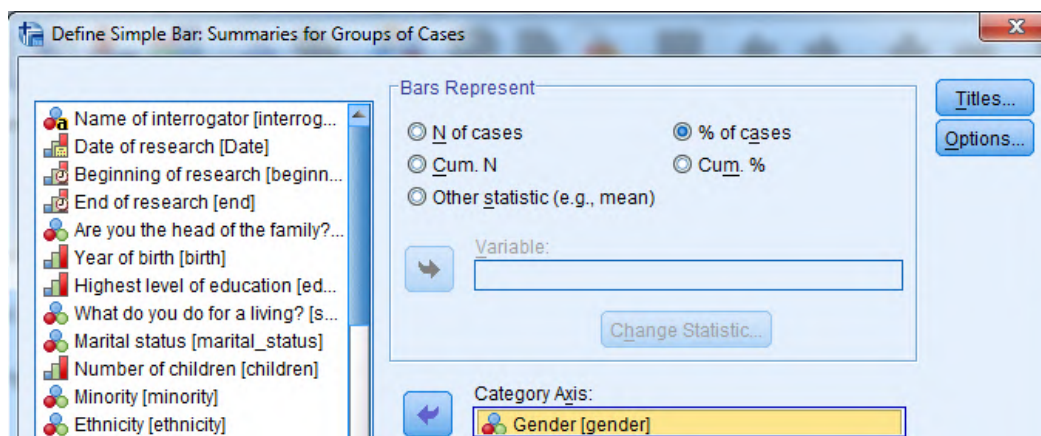


Figure 10 – Bar Charts Dialog Box

Graph is stored in the Output window. To edit a Legacy Dialogs graph, double click on the graph, a Chart Editor window appears. Alternatively, you can also right-click on the chart and select Edit Content and then select 'In separate window'.²

Optionally, you can change the proportions of the chart: e.g. colour, depth and angle (3D), width of bars, etc. For attaching the proportion values to the chart, select Data Label Mode, as on the Figure below.

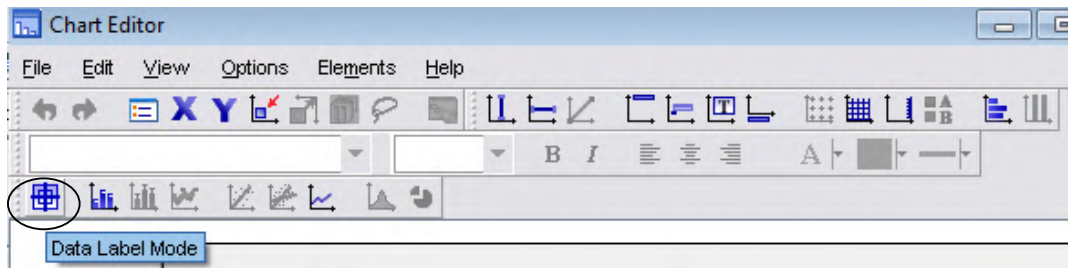


Figure 11 – Chart Editor

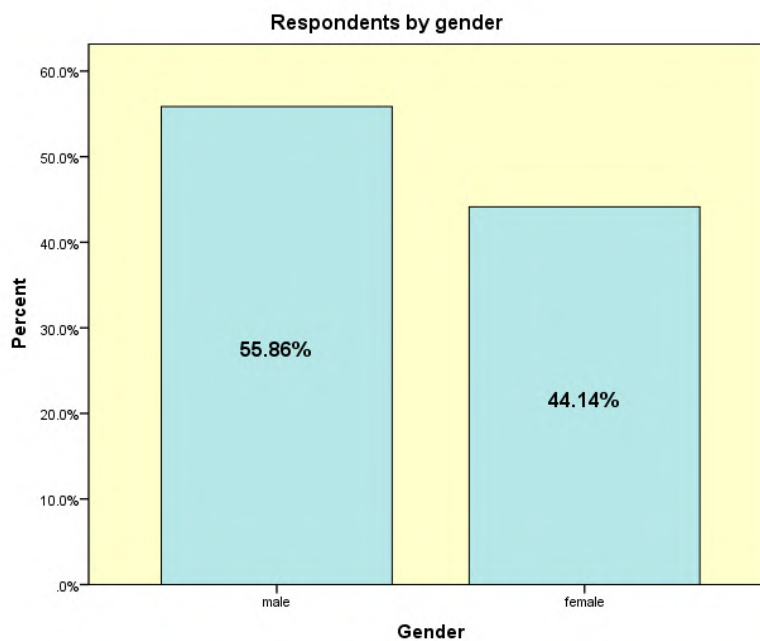


Figure 12 – Bar Chart

² SPSS Online Training Workshop, Central Michigan University (accessed: 05-01-2011)

Example 2 Transform the bar chart into a pie chart!

In order to transform a chart, click the previously edited bar chart in the Chart Editor and select the Properties from the menus: *Edit / Properties / Variables / Element Type...*

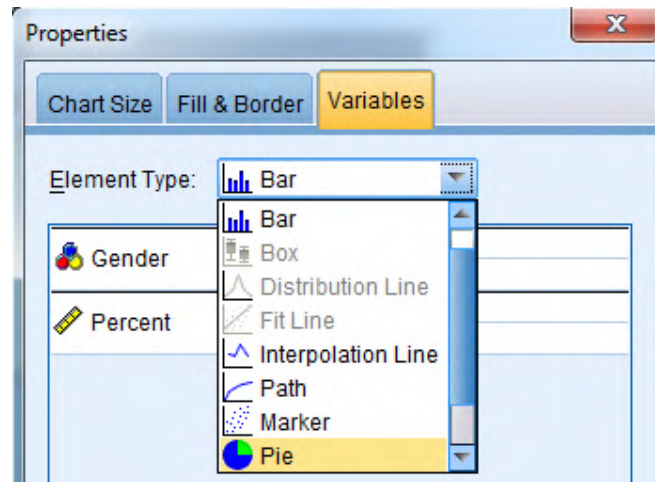


Figure 13 – Chart Editor Properties Dialog Box

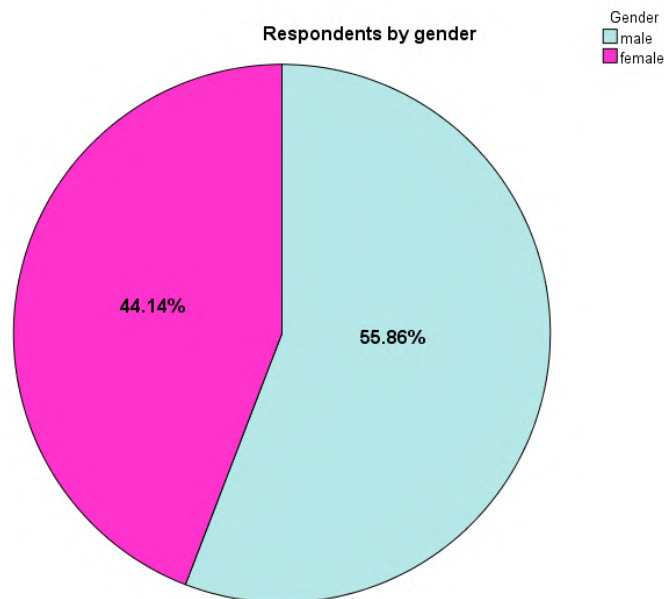


Figure 14 – Pie Chart

Example 3

Create a column diagram about the proportion of respondents grouped by *education level* stacked by *gender*! Embellish the graph!

The only difference between Example 1 and 3 is that now we should create stacked bar chart. Bar chart can be obtained by clicking on *Graphs* menu and selecting *Legacy Dialogs / Bar Charts* and then selecting the *stacked* type of bar chart (as on Figure 13). Then define category axis (Education level) and stacks (Gender).

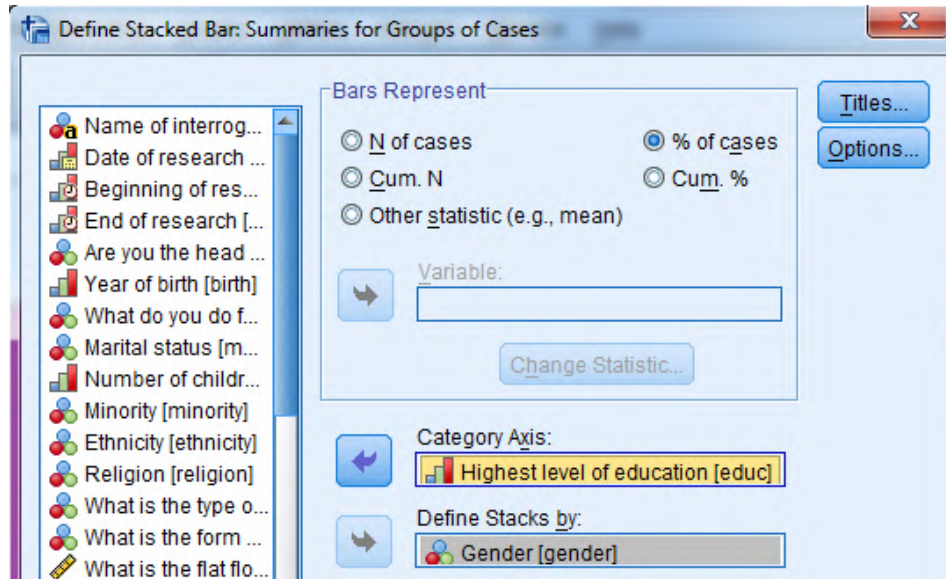


Figure 15 – Stacked Bar Chart Dialog Box

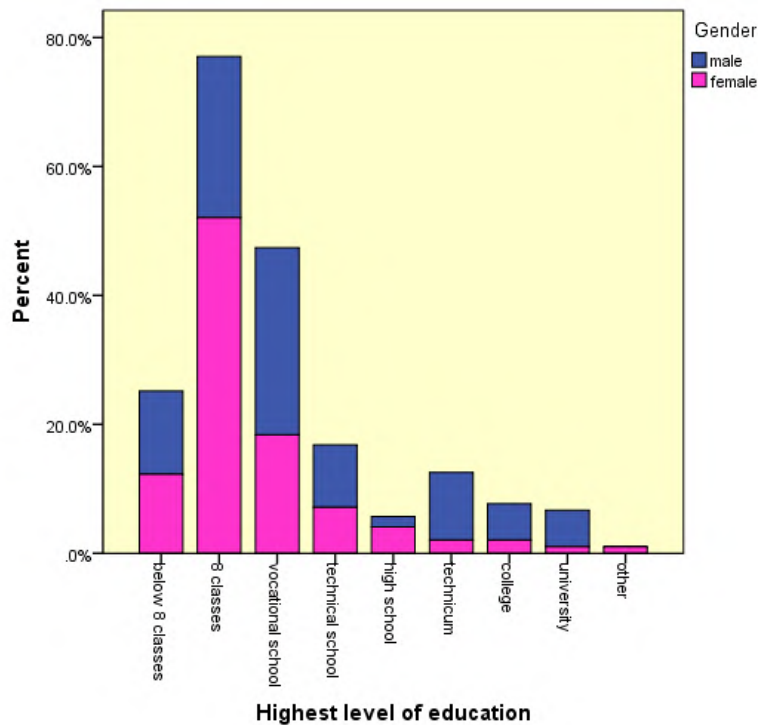


Figure 16 – Stacked Bar Chart

Example 4

Create a scatter plot of *average income* and *total expenditure* of households if you set markers by the type of heating (*heating_bin*)! Embellish the graph!

Scatter plot can be obtained by clicking on *Graphs / Legacy Dialogs / Scatter/Dot...*, and then the following box will appear. Simple scatter plot should be chosen.

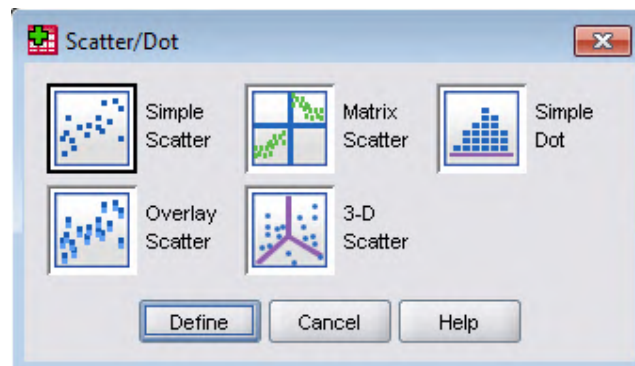


Figure 17 – Scatter / Dot Dialog Box

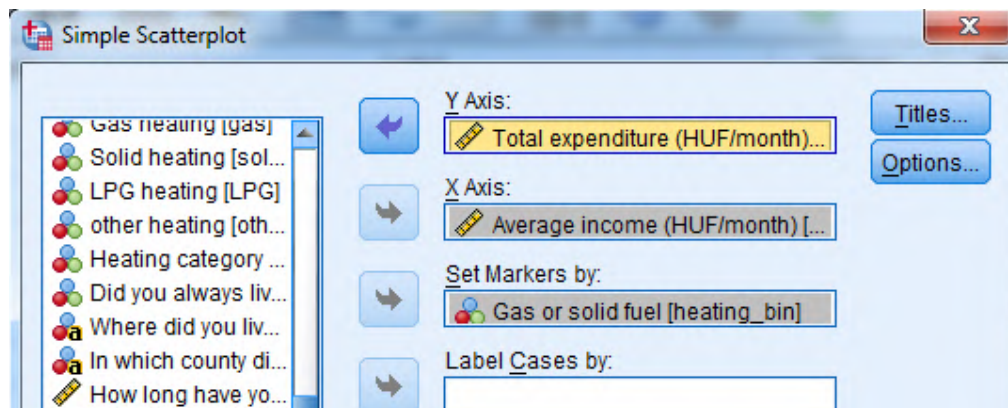


Figure 18 – Simple Scatter Plot Dialog Box

First we should define the axes (x: average income; y: total expenditure), then set markers by the type of heating (*heating_bin*). Optionally, we can label cases by a variable. If we have the names of respondents, that would be the label. Using the Chart Editor, we also can embellish the chart (change the colour or the type of markers).

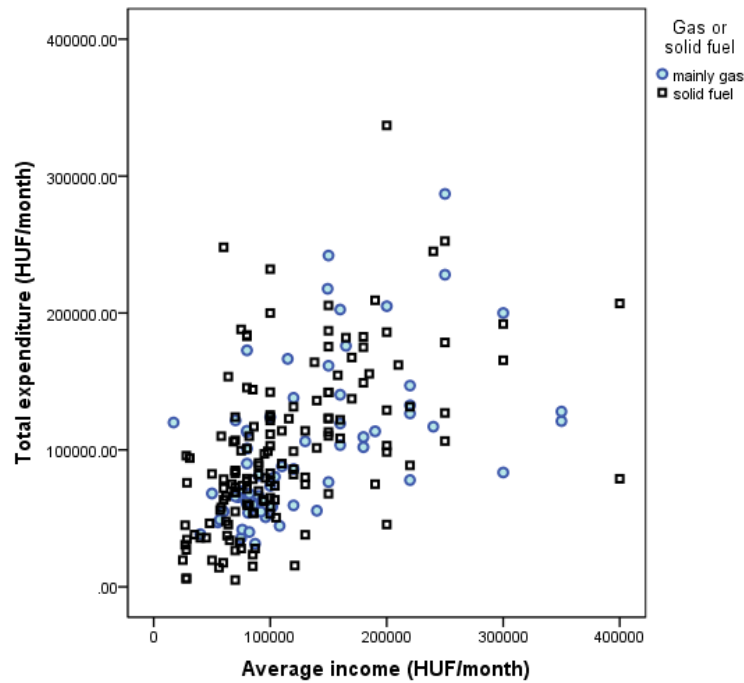


Figure 19 – Simple Scatter Plot

Example 5 Define a horizontal box plot of *total expenditure*! Embellish the graph!

First of all, we should define, what a box plot means. The box plot is a set of summary measures of distributions, like median, lower quartile, upper quartile, the smallest and the largest observations, moreover, the asymmetry can be seen as well.

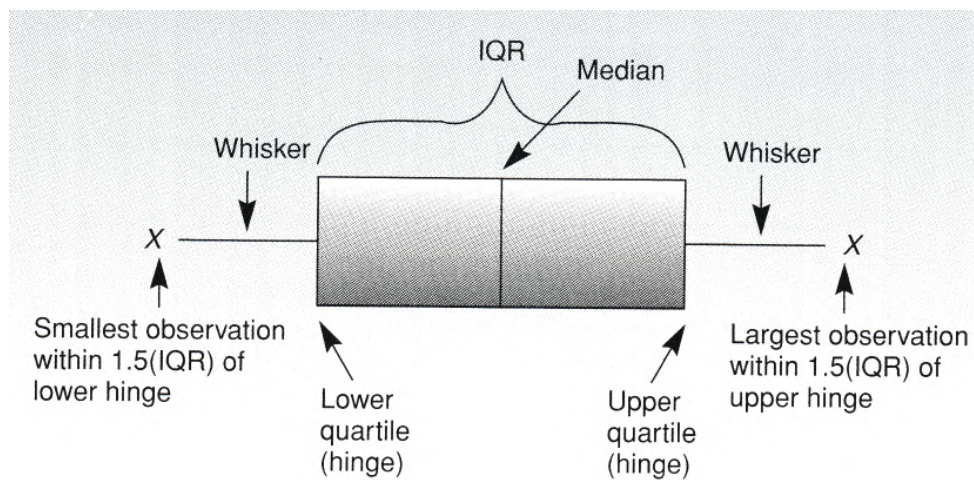


Figure 20 – Box Plot
Source: Aczel, 1996

Turning back to the exercise, we should create a box plot: *Graphs / Legacy Dialogs / Boxplot*. Choose simple chart to create a plot of one variable, and clustered for a comparison of variable types. Now we need a simple box plot of current salary, where data are summaries of separate variables.

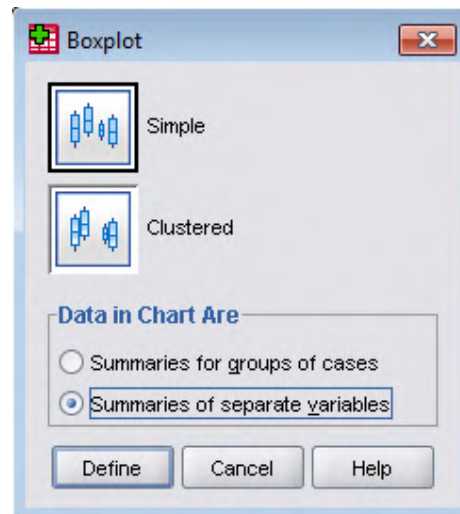


Figure 21 – Box Plot Dialog Box

For a horizontal box plot we need to transpose the chart: *Chart Editor / Options / Transpose Chart*.

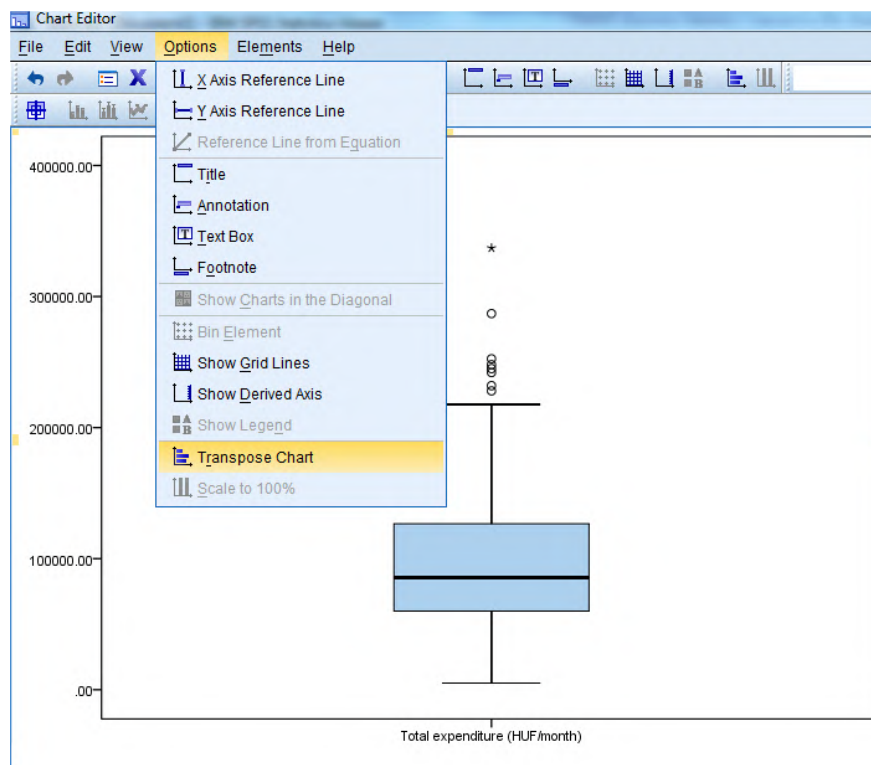


Figure 22 – Box Plot Chart Editor

Example 6

Define box plot of *total expenditure* of households categorized by the type of heating (*heating_bin*) clustered by *household clusters*! Embellish the graph!

For a categorized chart, choose the clustered box plot, where data in chart are summaries for groups of cases. The selected variable is the *total expenditure* and the type of heating (*heating_bin*) is on the category axis. The clusters are defined by *household clusters*.

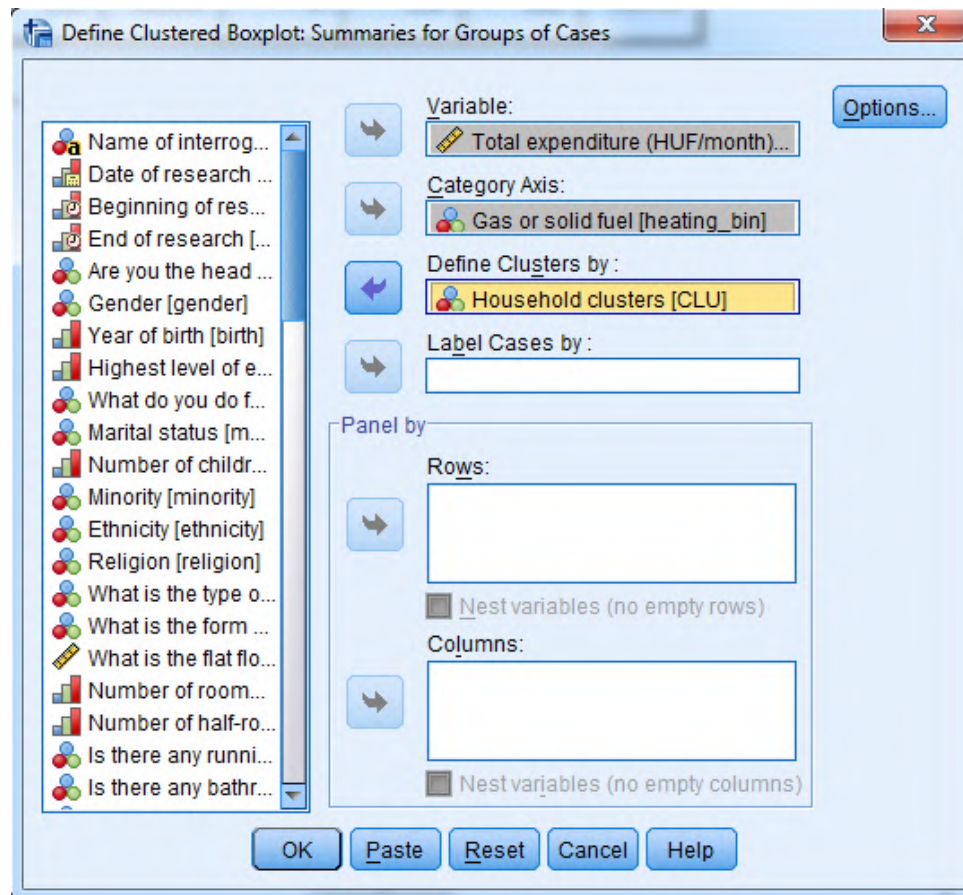


Figure 23 – Clustered Box Plot Chart Editor

The graph is edited by double clicking on the graph and double clicking on the part of graph which wanted to be edited. Then a *Properties* dialog box will appear to make changes. After making the changes, click on *Apply* to effect the changes and then close the dialog box.

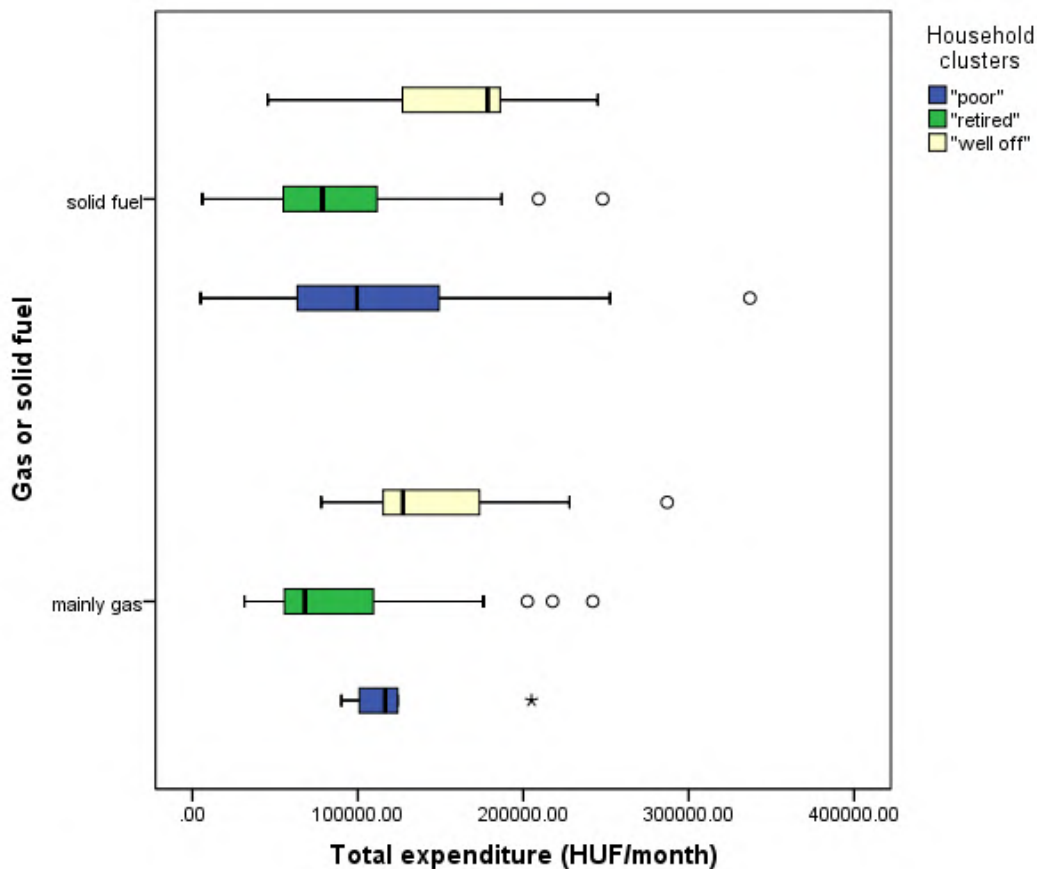


Figure 24 – Clustered Box Plot

4. CENTRAL TENDENCIES, MEASURES OF DISTRIBUTION, MEASURES OF ASYMMETRY

Example

Define the central tendencies, measures of distribution, measures of asymmetry and quartiles for total expenditure (HUF/month) of households!

These are measures of descriptive statistics, which are obtained by clicking on *Analyze* menu and selecting *Descriptive Statistics* then *Descriptive...*

Select the variable, which should be analyzed. If we would like, we can standardize variables as well, as it is shown on the figure below.

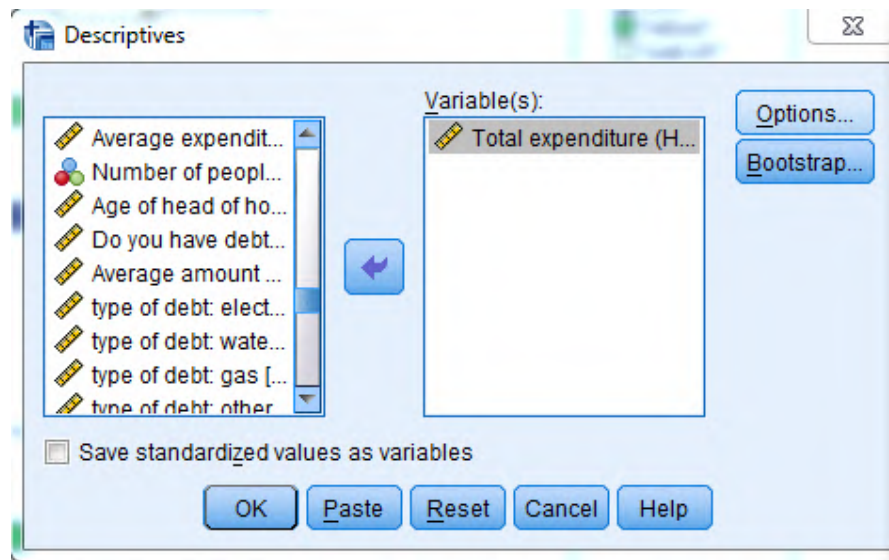


Figure 25 – Descriptives Dialog Box

Click on Options for optimal statistics.

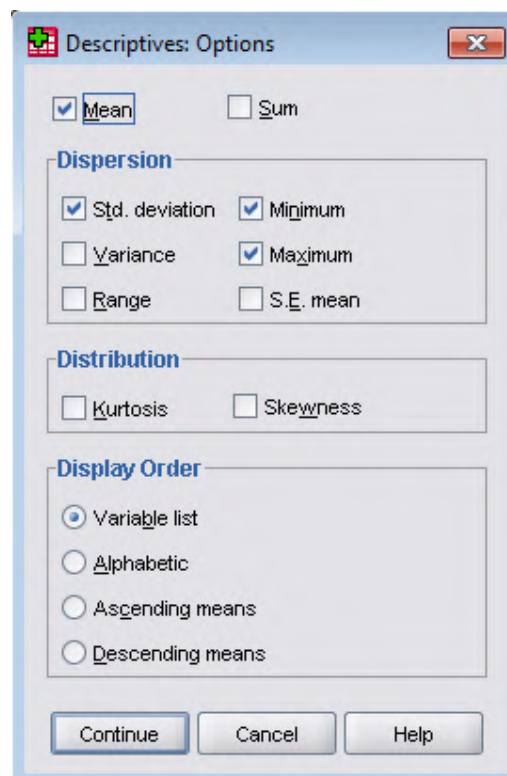


Figure 26 – Clustered Box Plot

Using this option we can define mean, standard deviation and the measure of asymmetry.

Alternatively, we reach descriptive statistics from *Analyze / Descriptive Statistics / Frequencies...* menu. After clicking on Statistics, the following box will appear:

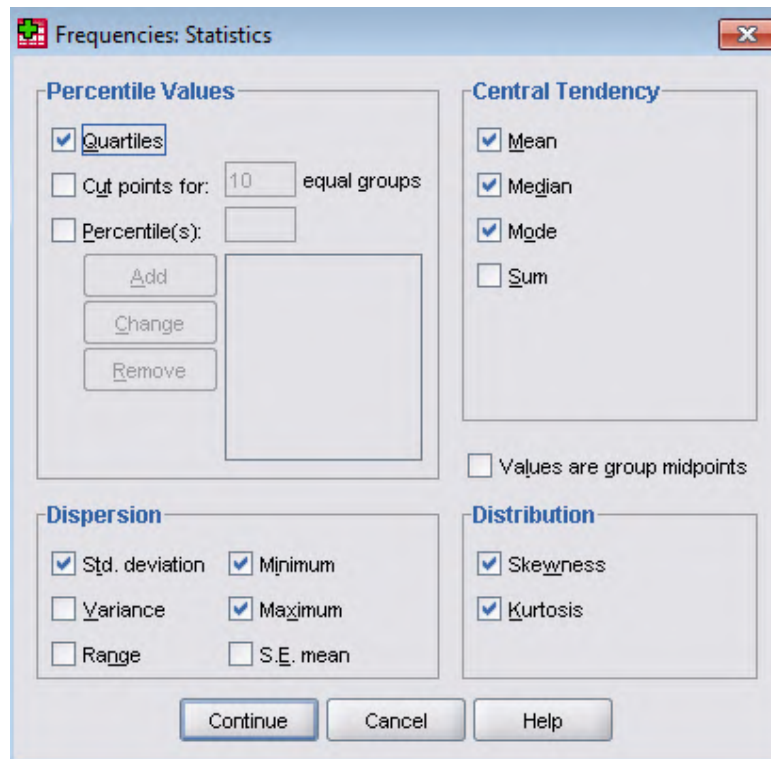


Figure 27 – Frequencies Statistics Dialog Box

The results are stored in the Output window.

Statistics		
Total expenditure (HUF/month)		
N	Valid	222
	Missing	0
Mean		99692.72
Median		85500.00
Mode		58500.00 ^a
Std. Deviation		57045.35
Skewness		1.02
Kurtosis		1.25
Minimum		5000.00
Maximum		337000.00
Percentiles	25	59875.00
	50	85500.00
	75	126700.00

a. Multiple modes exist. The smallest value is shown

The following are interpretations of figures:

- 222 households were examined in Csernely. (Number of cases)
- The average monthly expenditure is 99 692.72 HUF.
- Half of the households spend more than 85 500 HUF, the other half of them spend less. (85 500 HUF is the value above and below which half of the cases fall.)
- The most frequently occurring monthly expenditure is 58 500 HUF. Multiple mode exist and 58 500 HUF is the smallest value.
- The average dispersion around the mean is 57045.35 HUF.
- Long right tail asymmetry. (A distribution with a significant positive skewness has a long right tail.)
- Positive kurtosis indicates that the observations cluster more and have longer tails than those in the normal distribution.
- The lowest total expenditure of households is 5 000 HUF.
- The highest total expenditure of households is 337 000 HUF.
- 25% of households have lower monthly expenditure than 59 875 HUF and 75% have higher.
- 75% of households have lower monthly expenditure than 126 700 HUF and 25% have higher.

5. ESTIMATION AND HYPOTHESIS TESTING

Example 1 Define a 95% confidence interval for the *total expenditure*!

For defining a confidence interval, t-test is available in the SPSS by clicking on *Analyze / Compare Means / One Sample T Test...*

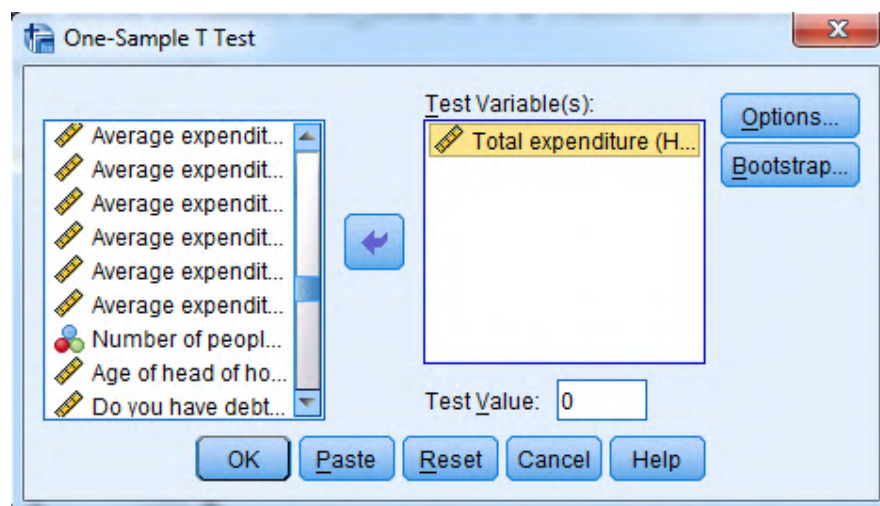


Figure 28 – One-Sample T Test Dialog Box for Estimation

In case of estimation, the test value should be zero.

Optionally, click on Options to control the confidence level. (The original setting is 95%). After clicking on OK, the results will be appeared in the Output window.

Test Value = 0					
t	df	Sig. (2-tailed)	Mean Difference	95% Confidence Interval of the Difference	
				Lower	Upper
26.039	221	0.000	99692.71622	92147.4136	107238.0189

The average total expenditure of households is between 92 147.4136 and 107 238.0189 HUF at 95% confidence level.

Example 2

Test the hypothesis that the *total expenditure* of households equals \$100 000. ($\alpha = 5\%$)

From the menu choose *Analyze / Compare Means / One Sample T Test...*

Enter the test value which each mean sample is compared. This is 100 000, as you can see below.

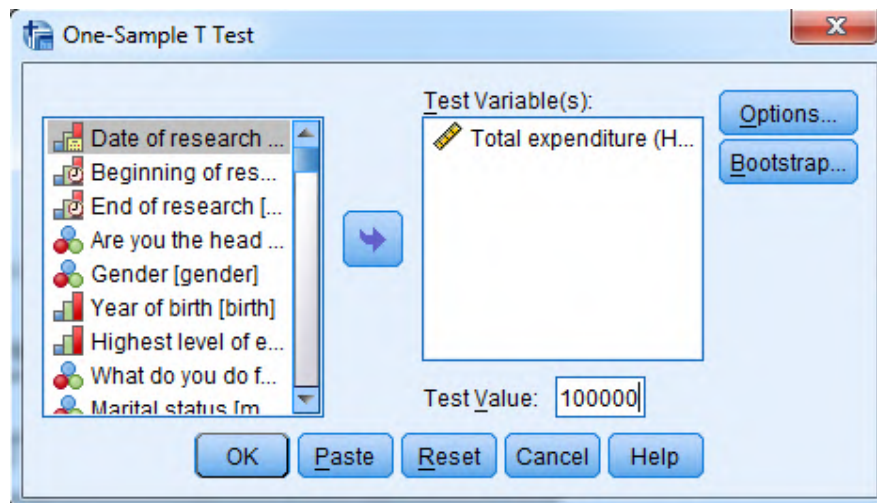


Figure 29 – One-Sample T Test Dialog Box for Hypothesis Testing

If the significance level is 5%, the confidence level will be 95%, which can be edited under the Options menu.

The results are the following:

Test Value = 100000					
t	df	Sig. (2-tailed)	Mean Difference	95% Confidence Interval of the Difference	
				Lower	Upper
-.080	221	.936	-307.28378	-7852.5864	7238.0189

If the p-value is less than 0.05, we reject the null. The p-value is 0.936, so we are basically declaring the null hypothesis to be true.

Example 3

Test the hypothesis that the *average expenditure on heating* of households heating with gas and households heating with solid fuels are equal! ($\alpha = 5\%$)

For comparing two independent means use independent t-test by clicking on *Analyze / Compare Means / Independent Samples T Test...*

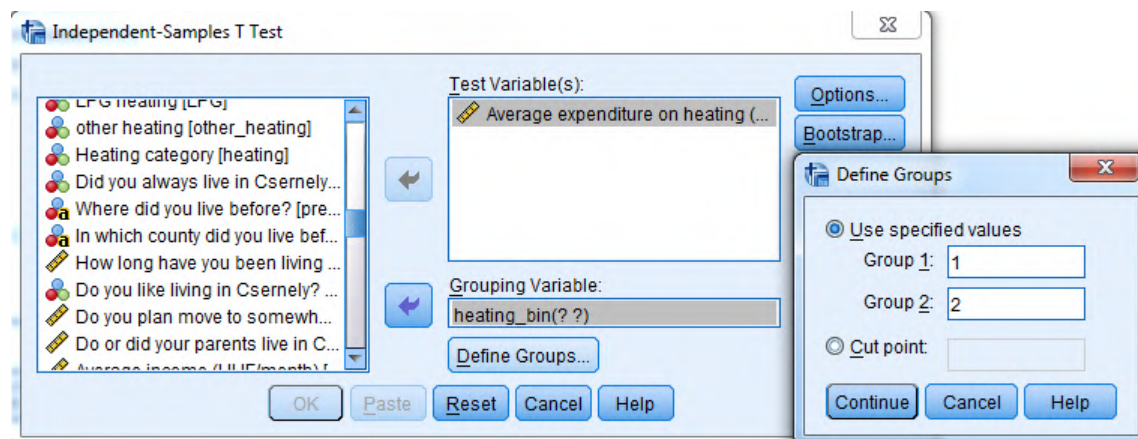


Figure 30 – Independent Samples T Test Dialog Box

Select the type of heating (*heating_bin*) as a grouping variable, where the groups are gas and solid fuels (coded as 1 and 2). Cases with any other values are excluded from the analysis. By clicking on Options, the confidence level can be changed.

If the population standard deviations are unknown, we have an assumption for equality of variances. Levene's test controls the equality of variances.

	Levene's Test for Equality of Variances		t-test for Equality of Means				
	F	Sig.	t	df	Sig. (2-tailed)	Mean Difference	Std. Error Difference
Equal variances assumed	5.012	0.026	-1.506	190	0.134	-5617.6	3729.59
Equal variances not assumed			-1.716	189.990	0.088	-5617.6	3273.48

We should analyze the first row, because equal variances assumed, because $n_{\text{gas}}=71$ and $n_{\text{solid_fuel}}=121$ (small sample size). When the F-value is large and the significance level of Levene's Test is small (smaller than say 0.1) the hypothesis of equal variances can be

rejected. The assumption is not significantly satisfied thus we cannot analyze the results of t-test. Anyway, we are basically declaring the alternative hypothesis to be true. Typically a conditional probability (critical significance level) of less than 0.1 or 0.05 is considered significant, thus the average expenditure on heating of households heating with gas or solid fuels are not equal.

Example 4 Nonparametric Tests – Hypothesis Testing for Distribution

Problem: Test the normality of average expenditure on heating! $\alpha=5\%$

Many parametric tests require normally distributed variables, thus we should test the hypotheses, whether the variable follows normal distribution or not.

H_0 : Normal distribution
 H_1 : Not normal distribution

Nonparametric hypothesis testing was applied to test for a normal distribution, that is why we will find this by clicking on *Analyze / Nonparametric tests / 1-Sample K-S Test* in the SPSS. One-Sample Kolmogorov – Smirnov Test procedure compares the observed cumulative distribution function for a variable with a specified theoretical distribution, which can be normal as well.

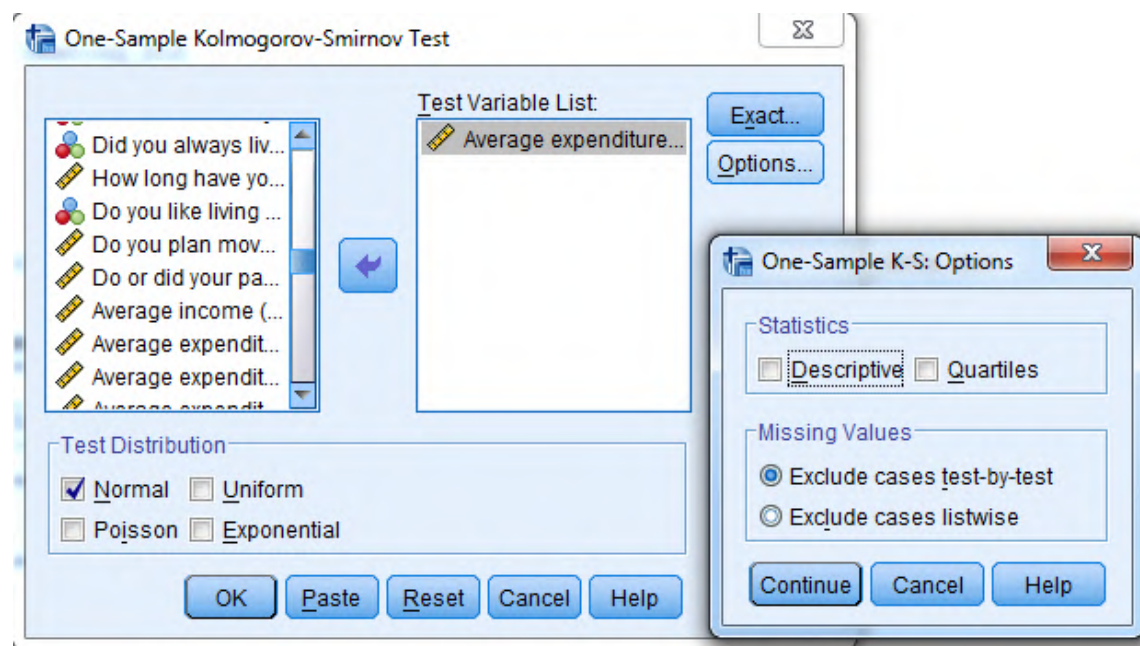


Figure 31 – One-Sample Kolmogorov–Smirnov Test Dialog Box

The test variable is now the average expenditure on heating and the test distribution is normal. If we want, we can generate descriptive statistics (usually including mean, standard deviation, sample size, minimum and maximum values, etc) by clicking on Options. It is good to know

them, because this procedure estimates the parameters from the sample, where the sample mean and standard deviation are the parameters for a normal distribution.

The following are the result in the Output View:

		Average expenditure on heating (HUF/month)
N		193
Normal Parameters ^{a,b}	Mean	31727.10
	Std. Deviation	25014.693
Most Extreme Differences	Absolute	.165
	Positive	.165
	Negative	-.122
Kolmogorov-Smirnov Z		2.290
Asymp. Sig. (2-tailed)		.000

a. Test distribution is Normal.

b. Calculated from data.

The p-value (asymptotic significance) tells you the probability of getting the results you got if the null were actually true. Thus the probability you would be in error if you rejected the null hypothesis is 0%. In other words, if the p-value is less than 0.05, you reject the normality assumption. So the average monthly expenditure on heating does not follow normal distribution.

Alternatively, there is another way of testing the normal distribution: using a histogram. The graph of the normal distribution depends on two factors - the mean and the standard deviation. The mean of the distribution determines the location of the center of the graph, and the standard deviation determines the height and width of the graph. When the standard deviation is large, the curve is short and wide; when the standard deviation is small, the curve is tall and narrow. All normal distributions look like a symmetric, bell-shaped curve.³

From the menus choose *Graphs / Legacy Dialogs / Histogram*, and then the following box will appear:

³ Statistics Tutorial: <http://stattrek.com> (accessed: 05-01-2011)

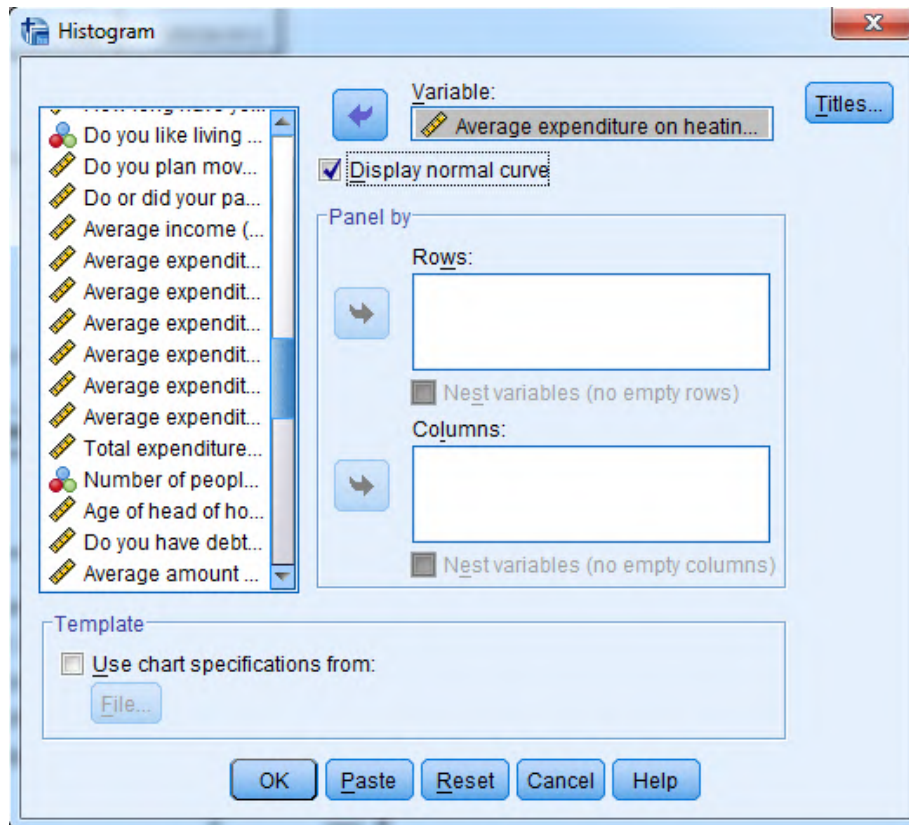


Figure 32 – Histogram Dialog Box

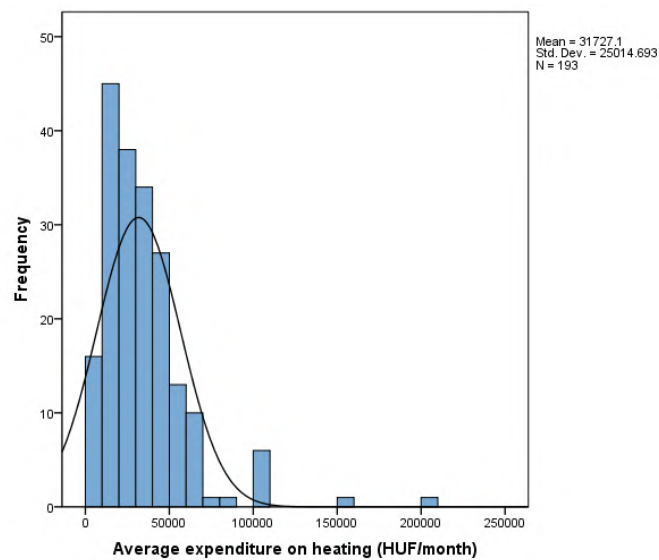


Figure 33 – Histogram

Therefore, the average monthly expenditure on heating is not normally distributed, because a truly normal curve is shaped like a bell that peaks in the middle and is perfectly symmetrical.

Example 5

Test the hypothesis, that *average expenditure on heating* of households heating with gas and households heating with solid fuels are equal, if we know that the average expenditure on heating does not follow normal distribution.

H_0 : The average expenditure on heating of households heating with gas and households heating with solid fuels are equal.

H_1 : The average expenditure on heating of households heating with gas and households heating with solid fuels are not equal.

In case of a non-normally distributed variable, we should use a nonparametric test for hypothesis testing. From the menus choose *Analyze / Nonparametric tests / 2-Independent Samples* and then the following dialog box will appear:

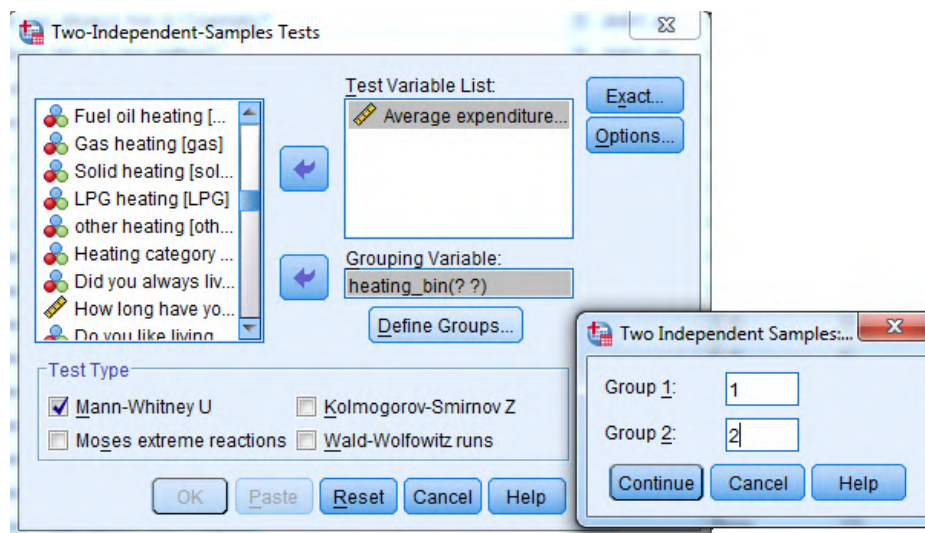


Figure 34 – Two-Independent-Samples Tests Dialog Box

Select the average expenditure on heating for test variable and heating category (*heating_bin*) for grouping variable. Click on Define Groups to split the file into two groups: Group 1: mainly gas (coded 1), Group 2: solid fuel (coded 2).

Mann-Whitney U Test is the most popular two-independent-samples test.

The following are the results from the Output window:

Test Statistics^a

	Average expenditure on heating (HUF/month)
Mann-Whitney U	4020.000
Wilcoxon W	6576.000
Z	-.743
Asymp. Sig. (2-tailed)	.458

a. Grouping Variable: Gas or solid fuel

The p-value is less than 0.1, so we reject the null hypothesis. The salary of clericals and managers are not equal.

6. STATISTICAL DEPENDENCE

Example 1 Create crosstabs from the type of heating (*heating_bin*) and household clusters!

Cross tabulation is the process of creating a contingency table from the multivariate frequency distribution of statistical variable. From the menus choose *Analyze / Descriptive Statistics / Crosstabs*.

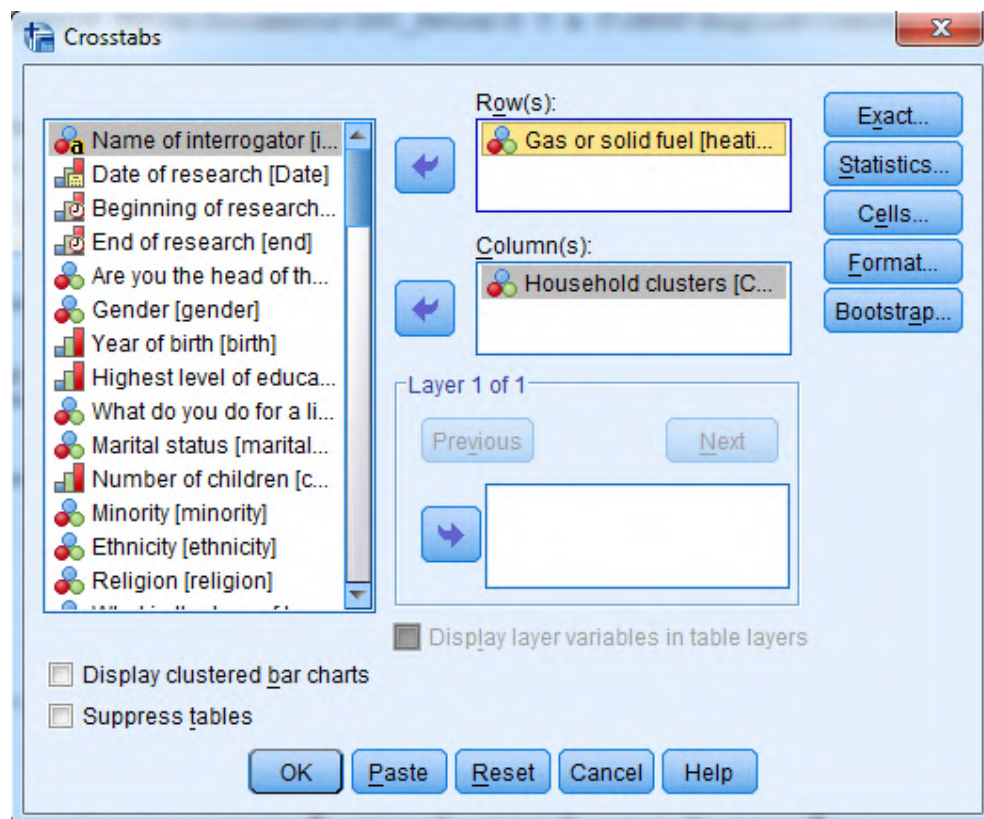


Figure 35 – Crosstabs Dialog Box

Select the row and column variable. It is up to you, which one is selected for row or column.

By clicking on cells, percentages or residuals can be displayed as well, as it is shown on the figure below.

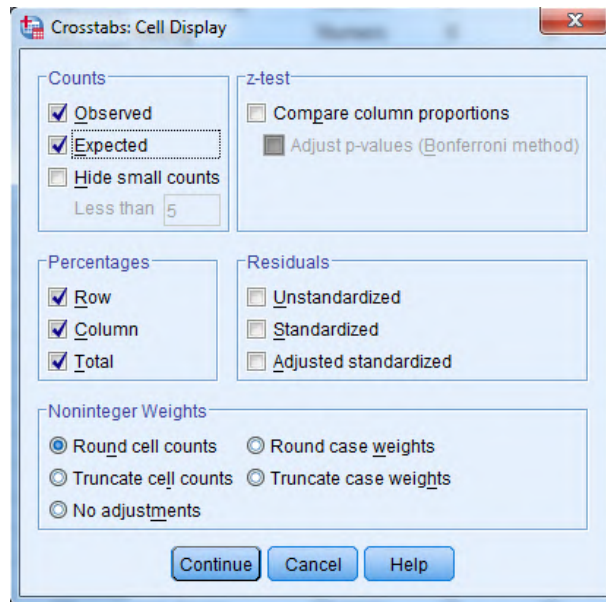


Figure 36 – Cell Display Dialog Box

Therefore, each cell of the statistical table can contain any combinations of counts, percentages (or residuals) selected. The observed counts are the frequencies (f_{ij}), while the expected counts are the frequencies for independence (f_{ij}^*).

$$f_{ij}^* = \frac{f_{i.} \cdot f_{.j}}{n}, \text{ where } f_{i.} \text{ and } f_{.j} \text{ are marginal frequencies.}$$

Gas or solid fuel * Household clusters Crosstabulation

			Household clusters			Total
			"poor"	"retired"	"well off"	
Gas or solid fuel	mainly gas	Count	6	53	12	71
		Expected Count	17.8	45.2	8.1	71.0
		% within Gas or solid fuel	8.5%	74.6%	16.9%	100.0%
		% within Household clusters	10.9%	37.9%	48.0%	32.3%
		% of Total	2.7%	24.1%	5.5%	32.3%
	solid fuel	Count	49	87	13	149
		Expected Count	37.3	94.8	16.9	149.0
		% within Gas or solid fuel	32.9%	58.4%	8.7%	100.0%
		% within Household clusters	89.1%	62.1%	52.0%	67.7%
		% of Total	22.3%	39.5%	5.9%	67.7%
Total	Count	55	140	25	220	
	Expected Count	55.0	140.0	25.0	220.0	
	% within Gas or solid fuel	25.0%	63.6%	11.4%	100.0%	
	% within Household clusters	100.0%	100.0%	100.0%	100.0%	
	% of Total	25.0%	63.6%	11.4%	100.0%	

Example 2 Is there any dependence between the type of heating (*heating_bin*) and household clusters?

Both variables are qualitative variables, thus we should calculate the measures of association for determining the strength of the relation.

Select *Analyze / Descriptive Statistics / Crosstabs*, as we did in Example 1. Add *heating_bin* to rows, and *household clusters* to columns. However, now click on Statistics if measures of association requested.

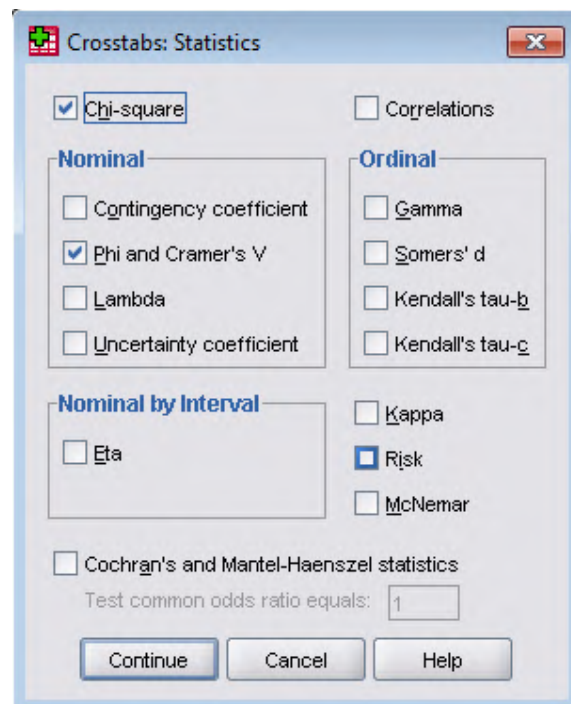


Figure 37 – Crosstabs Statistics Dialog Box

Chi-square should be selected to calculate the Pearson chi-square and the likelihood-ratio. Chi-square test is always a right tail test. Pearson chi-square statistic is used to test the hypothesis that the row and column variables are independent. Independence exists if the probability of their joint occurrence is equal to the product of their marginal probabilities.

H_0 : the two classification variables are **independent** of each other ($P_{ij} = P_i \cdot P_j$)

H_1 : the two classification variables are **NOT independent** ($H_1: \exists ij: P_{ij} \neq P_i \cdot P_j$)

The degrees of freedom are the degrees of freedom for the row variable times the degrees of freedom for the column variable. It is the product of the two degrees of freedom.

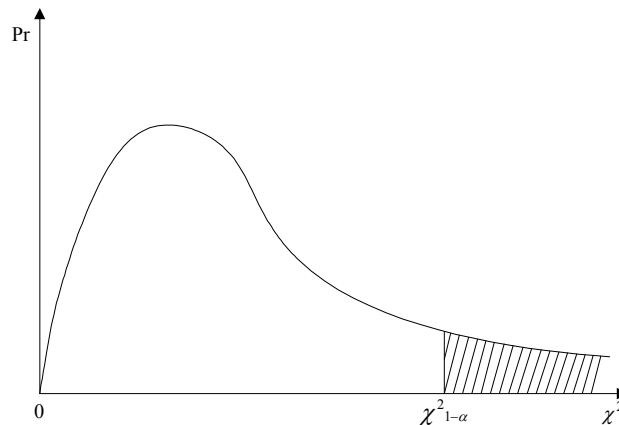


Figure 38 – Chi-square Distribution

For general purposes, the significance value is more important than the value of the statistic. Typically, a significance value of less than 0.05 is considered significant. For large sample sizes, the Pearson and Likelihood ratio statistics are equivalent.

For nominal data you can select Cramer’s V, which is the measure of association based on chi-square. If the Cramer’s measure is less than 0.3, the strength of dependence is weak. If it is between 0.3 and 0.7, the strength of dependence between variables is medium-strong. And if it is higher than 0.7, we say the dependence strong.

Chi-Square Tests

	Value	df	Asymp. Sig. (2-sided)
Pearson Chi-Square	16.311 ^a	2	.000
Likelihood Ratio	18.454	2	.000
Linear-by-Linear Association	14.754	1	.000
N of Valid Cases	220		

a. 0 cells (.0%) have expected count less than 5. The minimum expected count is 8.07.

According to chi-square statistic, the variables are not independent.

Symmetric Measures

	Value	Approx. Sig.
Nominal by Nominal Phi	.272	.000
Cramer's V	.272	.000
N of Valid Cases	220	

There is a weak dependence between the type of heating & household clusters. We can accept the statement at every significance level.

Example 3 Is there any dependence between the type of heating (*heating_bin*) and the average expenditures on heating?

The dependent variable, which is the current salary, is quantitative and the independent variable, gender is categorical. Thus it is a mixed dependence. To obtain the measures of mixed dependence, choose the *Analyze* menu then *Compare Means / Means*.

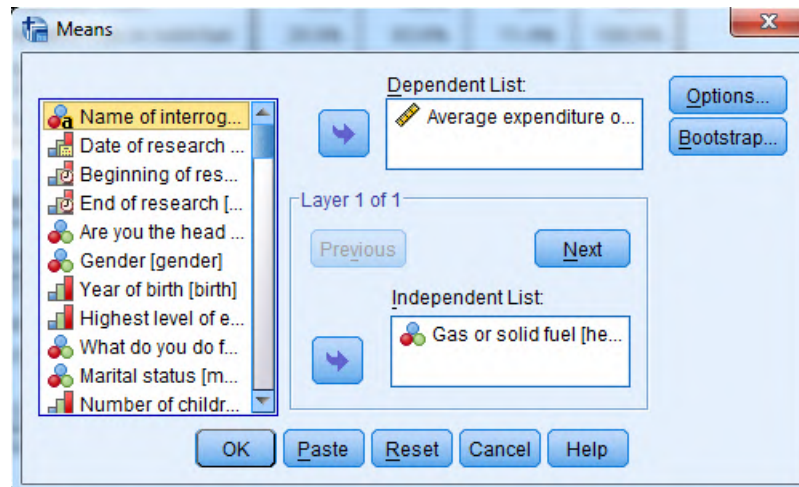


Figure 39 – Means Dialog Box

We can define the subgroup means and standard deviations by clicking on Options. The measures of mixed dependence can be found here as well.

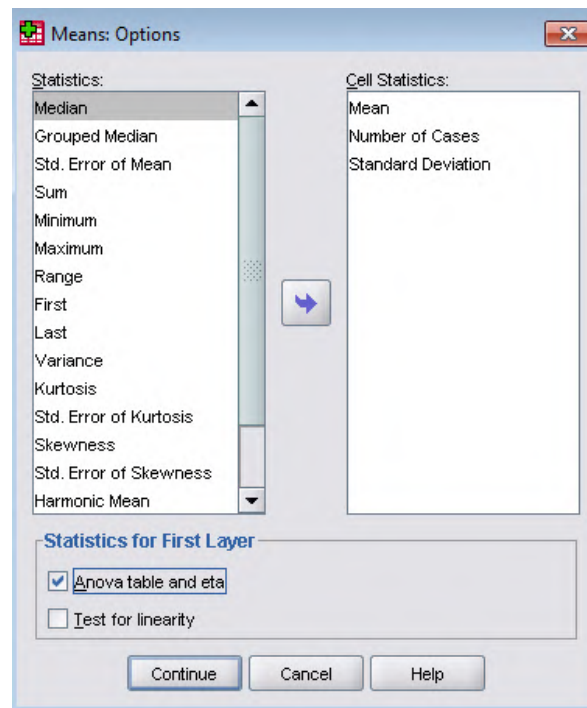


Figure 40 – Means Options Dialog Box

Because cell statistics were selected, the following table is displayed in the Output window.

Report

Average expenditure on heating (HUF/month)

Gas or solid fuel	Mean	N	Std. Deviation
mainly gas	28300.00	71	16660.861
solid fuel	33917.60	121	28697.218
Total	31840.26	192	25030.512

This table shows the central tendency & dispersion of the dependent variable (average expenditure on heating) grouped by the independent variable (*heating_bin*).

ANOVA table displays a one-way analysis-of-variance table to determine F-statistic. F-value is the ratio of two mean squares. A large F-value with small significance level indicates that the results probably are not due to random chance.

Thus we should take the probability (Sig.) into consideration, which shows that the relationship as strong as the one observed in the data would be present, if the null hypothesis were true. Typically, a value less than 0.05 is considered significant.

ANOVA Table

		Sum of Squares	df	Mean Square	F	Sig.
Average expenditure on heating (HUF/month) * Gas or solid fuel	Between Groups (Combined)	1.412E9	1	1.412E9	2.26	.134
	Within Groups	1.183E11	190	6.224E8		
	Total	1.197E11	191			

Now we can say that there is no significant relationship between the heating type and average expenditure on heating. (However, at a significance level higher than 13.4% we can accept the statement, that there is relation between the variables.) The degree of strength is shown below:

Measures of Association

	Eta	Eta Squared
Average expenditure on heating (HUF/month) * Gas or solid fuel	.109	.012

Eta is the so-called H-measure, which shows the degree of strength of relationship between variables. The value is higher than 0.3 but less than 0.7, thus the relationship between the heating type and the average expenditure on heating is weak. (Values close to 1 indicate a high degree of relationship). Eta squared is often called H²-measure, which is the proportion of variance in the dependent variable explained by differences among groups. Therefore 1.2% of variance in the average expenditure on heating is explained by the heating type.

7. CORRELATION AND LINEAR REGRESSION

Example 1 Is there any relationship between the total expenditure and average income?

Beginning salary and current salary are quantitative variables, so the relationship between them is a correlation, thus the correlation coefficient should be calculated.

From the menus choose *Analyze / Correlate / Bivariate*. For quantitative, normally distributed variables, choose the Pearson correlation coefficient. If the data are not normally distributed or have ordered categories, choose Kendall's tau-b or Spearman, which measure the relationship between rank orders. (SPSS 16.0 Tutorial)

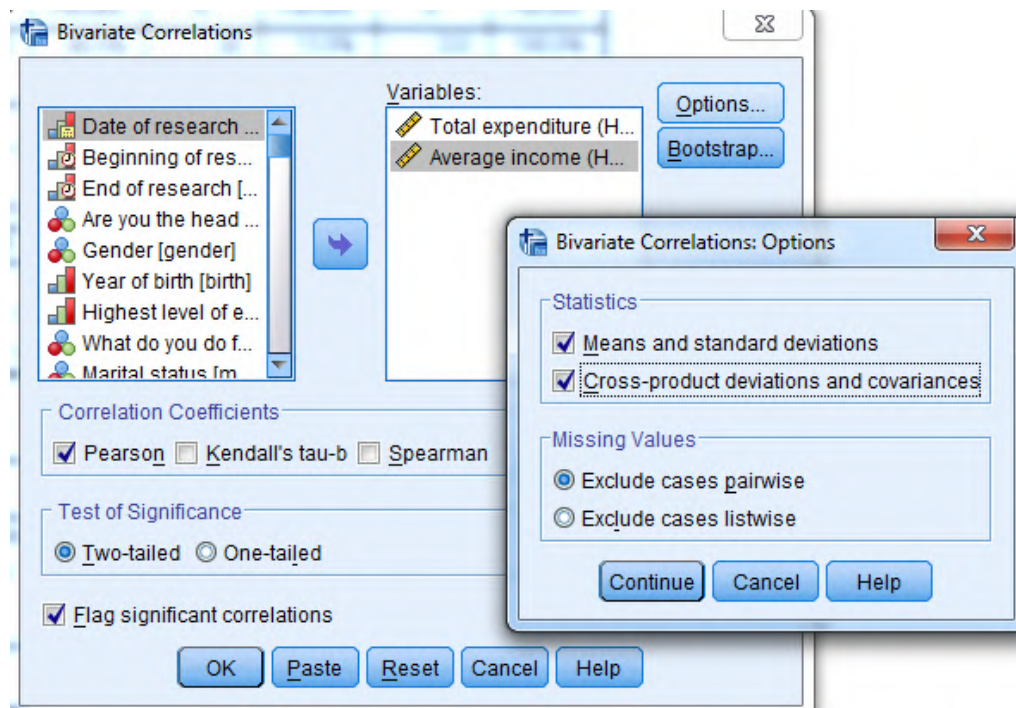


Figure 41 – Bivariate Correlations Dialog Box

Optionally, covariance can be displayed by clicking on Options. Covariance is an unstandardized measure of the relationship between two variables, equal to the cross-product deviation divided by $n-1$. It shows the direction of a relationship (positive or negative relationship).

		Total expenditure (HUF/month)	Average income (HUF/month)
Total expenditure (HUF/month)	<i>Pearson Correlation</i>	1	0.539(**)
	Sig. (2-tailed)		0.000
	<i>Covariance</i>	3254171744.937	2147784135.965
	N	222	217
Average income (HUF/month)	Pearson Correlation	0.539(**)	1
	Sig. (2-tailed)	0.000	
	Covariance	2147784135.965	4814606330.816
	N	217	217

There is a positive, medium-strong and significant (Sig.=0.000) relationship between the average expenditure and income.

Example 2 Determine a linear relation between the average monthly income and total expenditure!

Linear regression estimates the coefficients of the linear equation, involving an independent variable, that best predict the value of the dependent variable. Which variable has an influence on the other one: the average monthly income or the total expenditure? The average monthly income can have an influence on the total expenditure, thus the dependent variable is the total expenditure and the independent variable is the average monthly income.

To obtain a linear regression analysis, from menus choose *Analyze / Regression / Linear*.

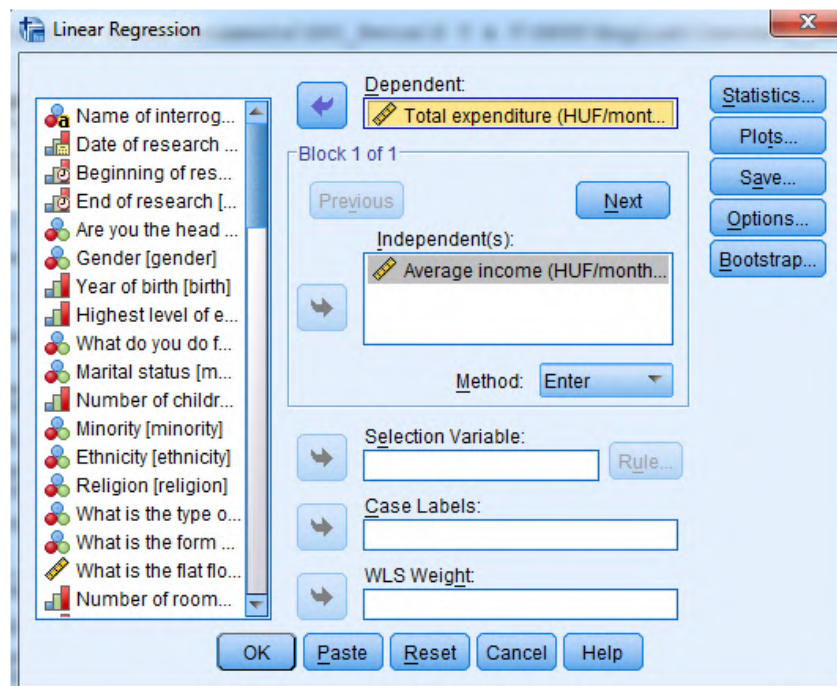


Figure 42 – Linear Regression Dialog Box

Optionally, we can display the covariance matrix and the matrix for correlation coefficients (again) by clicking on Statistics, as it is shown on the figure below.

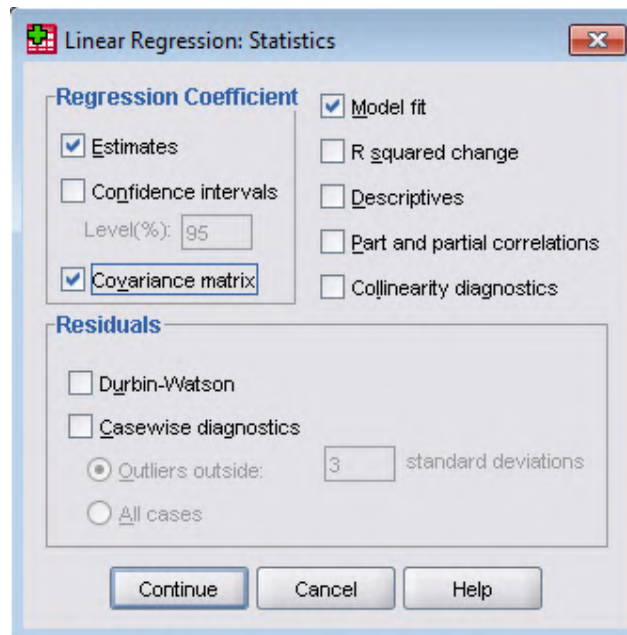


Figure 43 – Linear Regression: Statistics Dialog Box

ANOVA table is used for test the significance of the overall regression. If the significance level is close to zero (lower than 0.05), the regression is significant.

ANOVA table for bivariate regression model

Model	Sum of Squares	df	Mean Squares	F
Regression	$S_{\hat{y}} = \Sigma(\hat{y}_i - \bar{y})^2$	1	$S_{\hat{y}}$	$F = \frac{S_{\hat{y}}}{S_e / (n - 2)}$
Residual	$S_e = \Sigma(y_i - \hat{y})^2$	n-2	$s_e^2 = S_e / (n - 2)$	
Total	$S_y = \Sigma(y_i - \bar{y})^2$	n-1		

ANOVA^b

Model	Sum of Squares	df	Mean Square	F	Sig.
1 Regression	2.070E11	1	2.070E11	88.012	.000^a
Residual	5.056E11	215	2.351E9		
Total	7.125E11	216			

a. Predictors: (Constant), Average income (HUF/month)

b. Dependent Variable: Total expenditure (HUF/month)

For analyzing the coefficients of linear regression, take a look on the table below.

Coefficients^a

Model	Unstandardized Coefficients		Standardized Coefficients	t	Sig.
	B	Std. Error	Beta		
1 (Constant)	47355.059	6534.580		7.247	0.000
Average income (HUF/month)	0.446	.048	.539	9.381	0.000

The constant is the intercept point, when $x=0$. In general, it does not have a meaning, so we do not interpret the constant. However, it would be the average total expenditure (47355.059 HUF) in the case of zero average income.

The b_1 coefficient tells you how much the dependent variable is expected to increase (if the coefficient is positive) or decrease (if the coefficient is negative) when that independent variable increases by one. Therefore, when the average income increases by 1 HUF, the total expenditure is expected to increase by 0.446 HUF.

Moreover, it is good to interpret the results of t-test, which test the significance of the parameters. A lower p-value than 0.05 (5%) is the generally accepted to reject the null hypothesis of using non-significant values.

Example 3 Determine 95% confidence interval for the b_1 parameter!

Analyze / Regression / Linear / Statistics

For displaying the confidence interval of b coefficients, click on the box next to 'Confidence intervals' and change the confidence level, if it is necessarily.

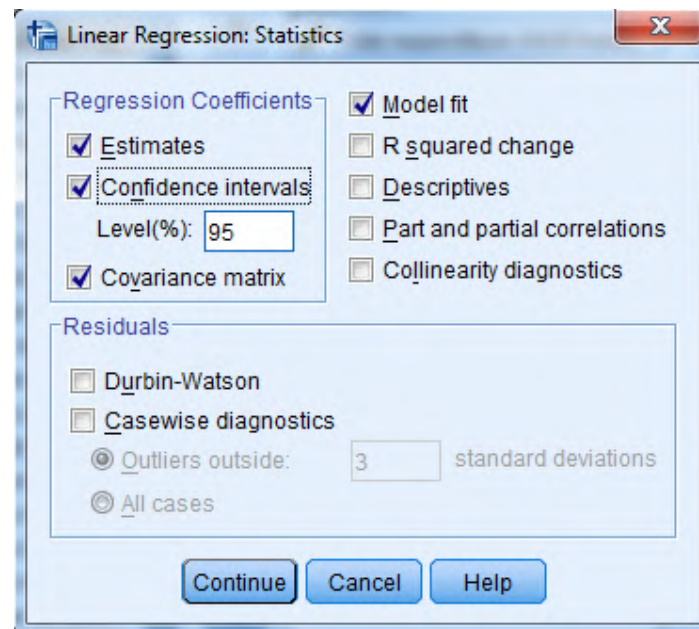


Figure 44 – Linear Regression Dialog Box: Statistics for Estimating the Coefficients

The following contain the confidence interval for b coefficient at 95% confidence level.

Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.	95.0% Confidence Interval for B	
		B	Std. Error	Beta			Lower Bound	Upper Bound
1	(Constant)	47355.059	6534.580		7.247	.000	34475.016	60235.101
	Average income (HUF/month)	.446	.048	.539	9.381	.000	.352	.540

a. Dependent Variable: Total expenditure (HUF/month)

The b_1 coefficient is between 0.352 and 0.540 at a 95% confidence level.

Example 4 Determine 95% confidence interval for the dependent variable (y: total expenditure)!

Analyze / Regression / Linear / Save

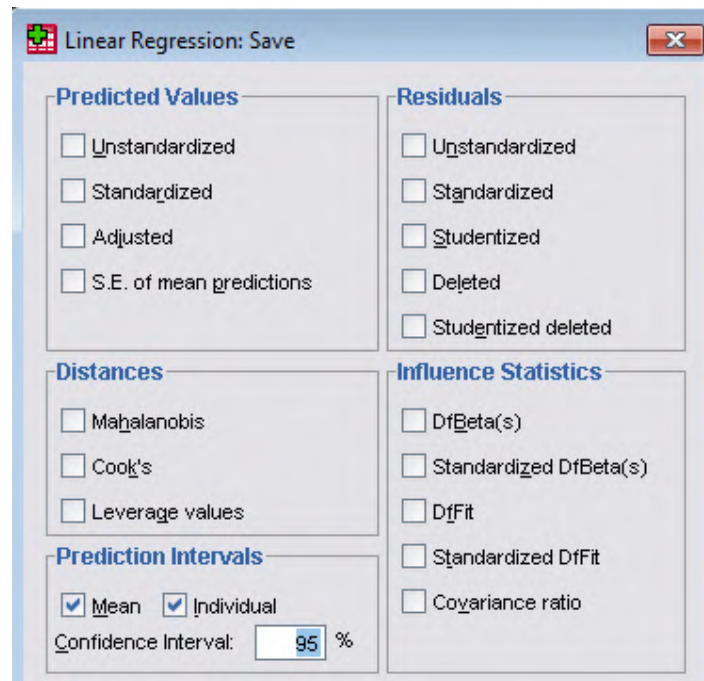


Figure 45 – Linear Regression: Save for Prediction Intervals

To display the upper and lower bounds for mean or individual prediction interval, click on the box of Mean or Individual. Enter 95% confidence interval.

Residuals Statistics^a

	Minimum	Maximum	Mean	Std. Deviation	N
Predicted Value	54 938.72	225 794.08	100 312.41	30 953.53	217
Std. Predicted Value	-1.466	4.054	.000	1.000	217
Standard Error of Predicted Value	3292.386	13 774.578	4378.560	1584.899	217
Adjusted Predicted Value	53 977.68	238 686.89	100 414.08	31 347.65	217

a. Dependent Variable: Total expenditure (HUF/month)

Therefore, the current salary of employees is between 54 938.72 and 225 794.08 at a 95% confidence level.

8. MULTIPLE CORRELATION AND LINEAR REGRESSION

Example 1

Is there any relation between the average income, total expenditure and number of people living in the household!

Analyze / Correlate / Bivariate / Pearson Correlation Coefficient

Variables: average income, total expenditure, number of people living in the household

Correlations

		Average income	Total expenditure	Number of people living in the household
Average income	Pearson Correlation	1	0.539**	0.293**
	Sig. (2-tailed)		0.000	0.034
Total expenditure	Pearson Correlation	0.539**	1	0.373**
	Sig. (2-tailed)	0.000		0.000
Number of people living in the household	Pearson Correlation	.293**	0.373**	1
	Sig. (2-tailed)	0.034	0.000	

** . Correlation is significant at the 0.01 level (2-tailed).

* . Correlation is significant at the 0.05 level (2-tailed).

$r_{\text{average income, total expenditure}} = 0.539$ – Positive, medium-strong, significant relationship

$r_{\text{average income, number of people}} = 0.293$ – Positive, weak, significant relationship

$r_{\text{total expenditure, number of people}} = 0.373$ – Positive, medium-strong, significant relationship

Example 2

Determine a linear relation between the average income, total expenditure and number of people living in the household!

Analyze / Regression / Linear / Dependent: Total expenditure; Independents: Average income and Number of people living in the household.

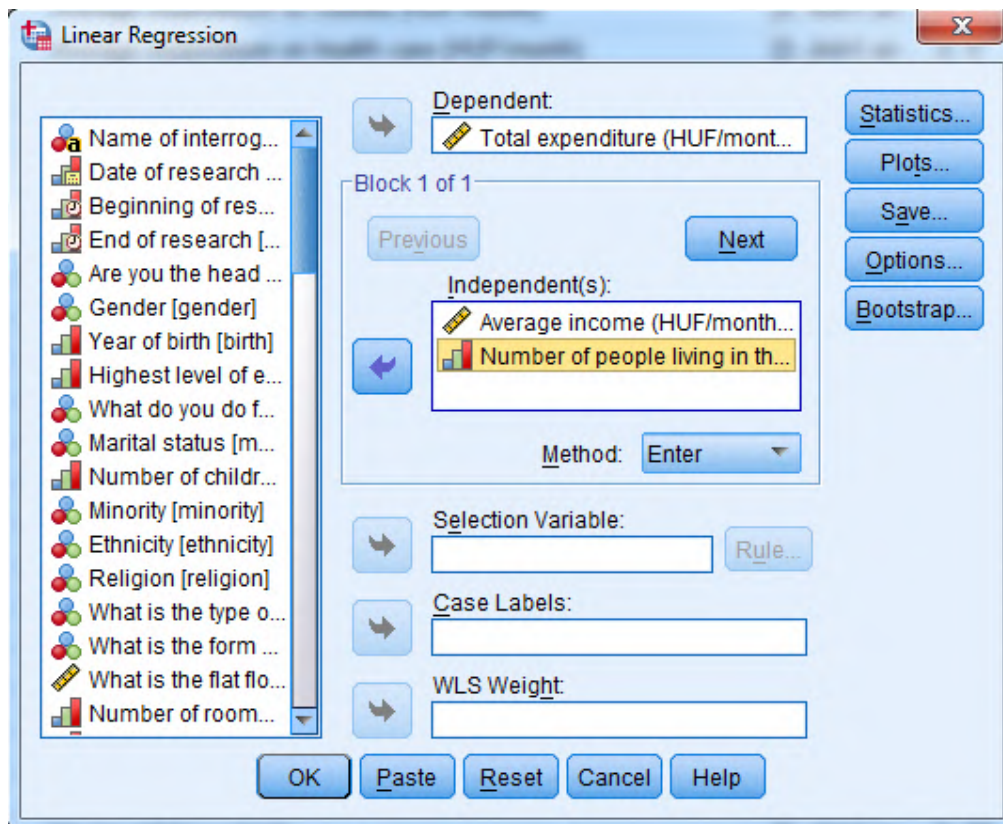


Figure 46 – Multiple Linear Regression

A multiple linear regression has more conditions:

- Assumptions for error term:
 - Normally distributed error;
 - Expected value of error= 0; $E(\epsilon)=0$;
 - The variance is the same for all observations (Homoscedasticity);
 - Uncorrelated across observations (no autocorrelation).
- Assumptions for the Independent Variables:
 - The independent variable is not random;
 - No multicollinearity (the predictors should not correlate).

These assumptions can be tested in the SPSS; however this is the topic of MSc courses that is why we do not show them.

The following are the results of linear regression:

Model Summary^b

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate	Durbin-Watson
1	0.581^a	0.338	0.332	46 948.77	1.780

a. Predictors: (Constant), Average income (HUF/month), Number of people living in the household

b. Dependent Variable: Total expenditure (HUF/month)

R (coefficient of correlation): there is a medium-strong relationship between the dependent variable (total expenditure) and the independent variables (average income & number of people living in the household).

R^2 (coefficient of determination): 33.8% of variance in the total expenditure explained by average income & number of people living in the household.

\bar{R}^2 : 33.2% of variance in the total expenditure explained by average income & number of people living in the household after a correction by the sample size and number of parameters. It corrects R^2 to more closely reflect the goodness-of-fit of the model in the population.

The following are the analysis of variance table for multiple regression model, which is to test the significance of the overall regression:

ANOVA table for multiple regression model

Model	Sum of Squares	df	Mean Squares	F
Regression	$S_{\hat{y}} = \Sigma(\hat{y}_i - \bar{y})^2$	p^4	$S_{\hat{y}}$	$F = \frac{S_{\hat{y}}}{S_e/(n-p-1)}$
Residual	$S_e = \Sigma(y_i - \hat{y})^2$	$n-p-1$	$s_e^2 = S_e/(n-p-1)$	
Total	$S_y = \Sigma(y_i - \bar{y})^2$	$n-1$		

ANOVA^b

Model	Sum of Squares	df	Mean Square	F	Sig.
1 Regression	2.408E11	2	1.204E11	54.627	0.000^a
Residual	4.717E11	214	2.204E9		
Total	7.125E11	216			

a. Predictors: (Constant), Average income (HUF/month), Number of people living in the household

b. Dependent Variable: Total expenditure (HUF/month)

The table above shows, that the multiple regression model is significant.

⁴ p: number of independent variables

Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.	95.0% Confidence Interval for B		Collinearity Statistics	
		B	Std. Error	Beta			Lower Bound	Upper Bound	Tolerance	VIF
1	(Constant)	32105.279	7427.351		4.323	.000	17465.143	46745.415		
	Average income (HUF/month)	.391	.048	.472	8.116	.000	.296	.486	.914	1.094
	Number of people living in the household	8334.457	2126.456	.228	3.919	.000	4142.976	12525.938	.914	1.094

a. Dependent Variable: Total expenditure (HUF/month)

Moreover, we have to give interpretations for the parameters.

b_0 : not analyzed (generally, it is the mean for the response when all of the independent variables (x) take on the value 0.)

b_1 : the total expenditure is expected to increase by 0.391 when the average income increases by 1 HUF, holding all the other independent variables constant.

b_2 : the total expenditure is expected to increase by 8334.457 HUF when the number of people living in the household increases by 1 people, holding all the other independent variables constant.

9. CURVILINEAR REGRESSION

Example

Which regression model fit the most on the relation total expenditure and average income?

We should test the types of regression by curve estimation: *Analyze / Regression / Curve Estimation / Dependent: Total expenditure; Independent: Average income; Models: Linear, Power, Compound.*

There are three methods how to test the best fitting regression model:

- Plot model (select '*Plot Model*'): it plots the values of the dependent variable and each selected model against the independent variable. In this way we can determine which model to use. If the variables appear to be related linearly, use a simple linear regression model. If the variables are not linearly related, use a curvilinear regression. The most frequently used curvilinear regressions are the power and compound regressions.
- ANOVA table (select '*Display ANOVA table*'): displays a summary analysis of variance table for each selected model. When the significance level is lower than 0.05, the model is well-fitted.
- Model summary: R^2 shows the goodness-of-fit (in case of linear regression). The higher is the R^2 , the better is the model fitting on the data.

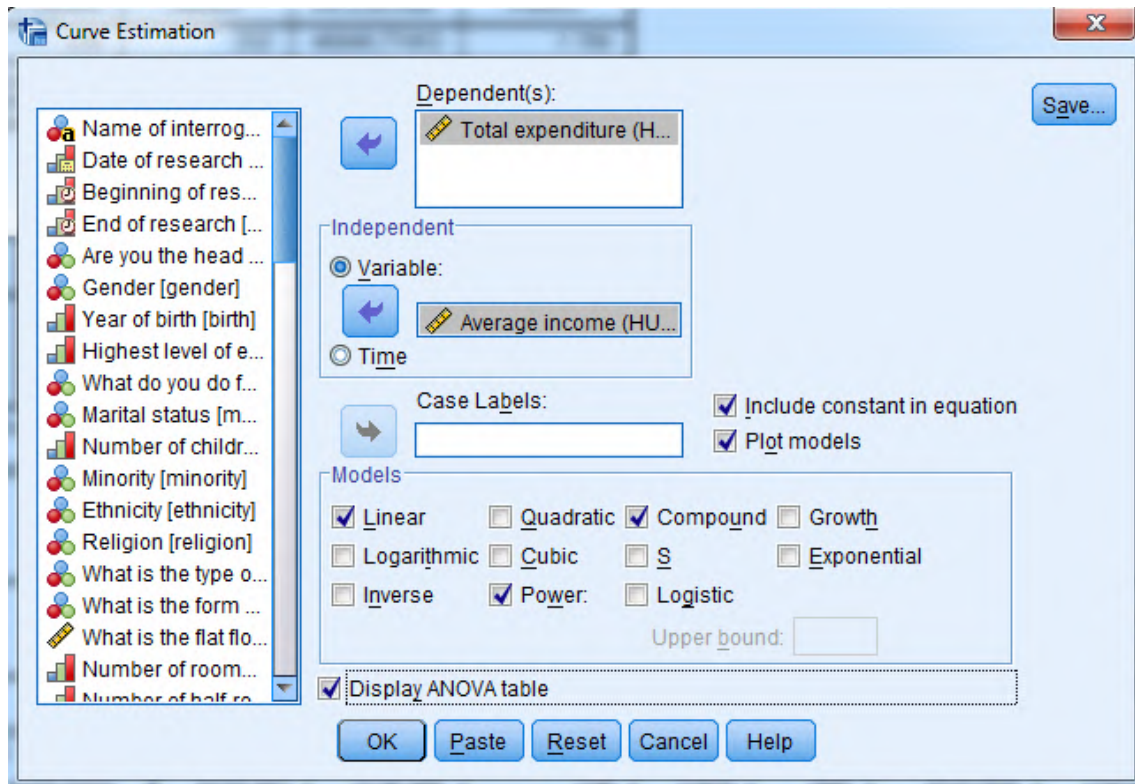


Figure 47 – Curve Estimation Dialog Box

The results are displayed in the Output window. The significance levels of F-tests are 0 in all cases. Thus we should compare the goodness-of-fit results.

Linear Regression Model Summary

R	R Square	Adjusted R Square	Std. Error of the Estimate
0.539	0.290	0.287	48 491.495

The independent variable is Average income (HUF/month).

Compound Regression Model Summary

R	R Square	Adjusted R Square	Std. Error of the Estimate
0.515	0.265	0.262	0.580

The independent variable is Average income (HUF/month).

Power Regression Model Summary

R	R Square	Adjusted R Square	Std. Error of the Estimate
0.574	0.330	0.326	0.554

The independent variable is Average income (HUF/month).

According to the goodness-of-fit results, the power regression model fit the most on the data. Alternatively, we can analyze the plot model. However, it is difficult to find the best fitting function according to a plot.

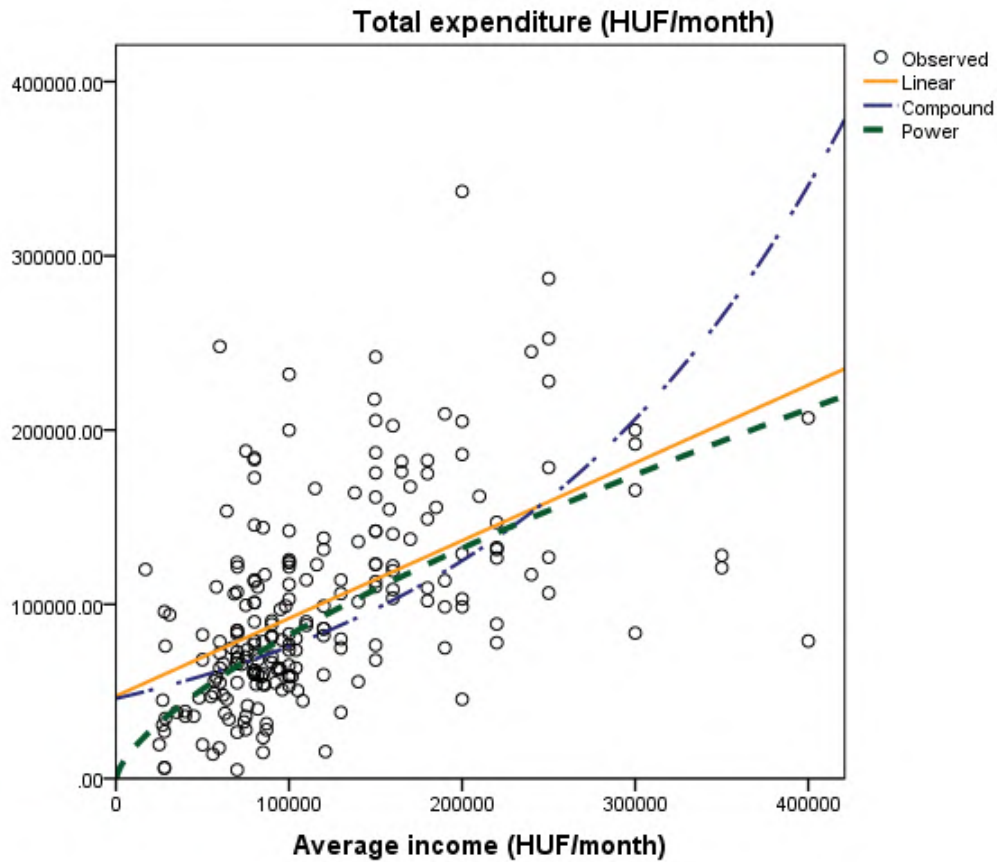


Figure 48 – Curve Fit

The data resemble a power regression function, thus we should give an interpretation of the parameters.

Coefficients

	Unstandardized Coefficients		Standardized Coefficients	t	Sig.
	B	Std. Error	Beta		
ln(Average income)	0.683	0.066	0.574	10.279	0.000
(Constant)	31.810	24.383		1.305	0.193

The dependent variable is ln(Total expenditures).

$$\hat{y} = a \cdot x^b = 31.810 \cdot x^{0.683}$$

a: 31.810 is the total expenditure of a household in Csernely, whose average income was 1 HUF.

b: If the average income increases by 1%, the total expenditures increases by 0.683%.

Note:

Power regression: $\hat{y} = a \cdot X^b$

- **a**: when $x=1$, $y=a$ ($a=10^{b0}$)
- **b**: when x increases by 1%, we expect y to change by $b\%$ ($b=b_1=E$).

Compound regression: $\hat{y} = a \cdot b^x$

- **a**: no analyzation
- **b**: when x increases by 1 unit, we expect y to be b times of the coefficient.

10. TIME SERIES ANALYZES

Example 1 The following are data about a population in 1994.

Month	Population (thousand persons)	Number of birth (persons)	Number of death (persons)
1	10 273	10 238	13 888
2	10 270	9 285	12 825
3	10 267	10 105	12 516
4	10 265	9 617	11 753
5	10 262	9 548	12 328
6	10 260	9 717	11 839
7	10 258	9 965	11 848
8	10 257	9 980	11 722
9	10 256	9 844	10 968
10	10 252	9 021	12 542
11	10 249	8 740	11 743
12	10 246	9 538	12 917

Source: HCSO

Problem:

- Create an SPSS data set!

Variable View:

	Name	Type	Width	Decimals	Label	Values	Missing	Columns	Align	Measure
1	month	Numeric	8	0	1994	None	None	8	Right	Scale
2	population	Numeric	8	0	population (th p)	None	None	8	Right	Scale
3	birth	Numeric	8	0	number of birth (person)	None	None	8	Right	Scale
4	death	Numeric	8	0	number of death (person)	None	None	8	Right	Scale

Data View:

	month	population	birth	death
1	1	10273	10238	13888
2	2	10270	9285	12825
3	3	10267	10105	12516
4	4	10265	9617	11753
5	5	10262	9548	12328
6	6	10260	9717	11839
7	7	10258	9965	11848
8	8	10257	9980	11722
9	9	10256	9844	10968
10	10	10252	9021	12542
11	11	10249	8740	11743
12	12	10246	9538	12917

a) Create a sequence chart about the number of birth and death!

From the menus choose: *Analyze / Forecasting / Sequence Charts* and then the following box will appear.

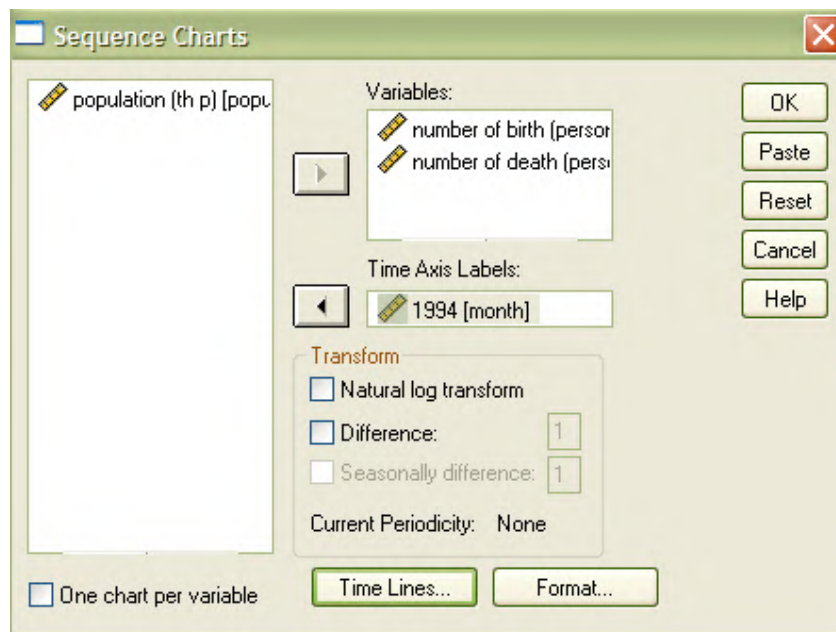


Figure 49 – Sequence Charts Dialog Box

We should define the variables: number of birth & number of death. Time axis labels the month.

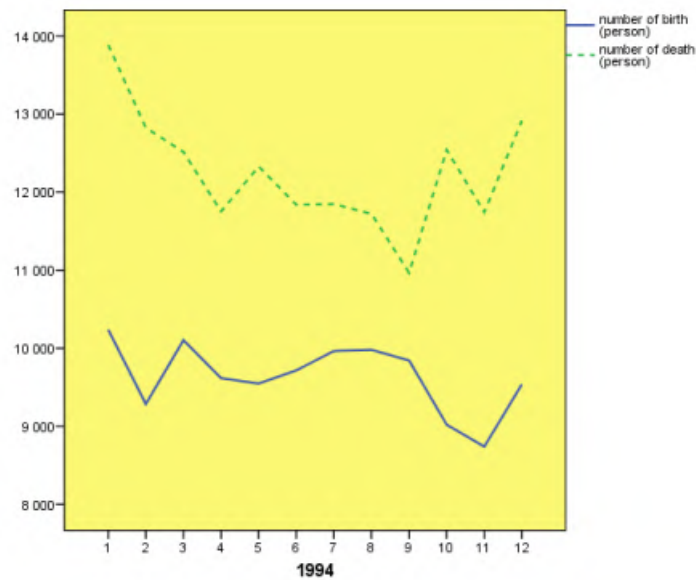


Figure 50 – Sequence Chart

b) Create a linear trend model for the number of birth!

From the menus choose: *Analyze / Regression / Curve Estimation*. The dependent variable is the number of birth; however, select *Time* for independent variable.

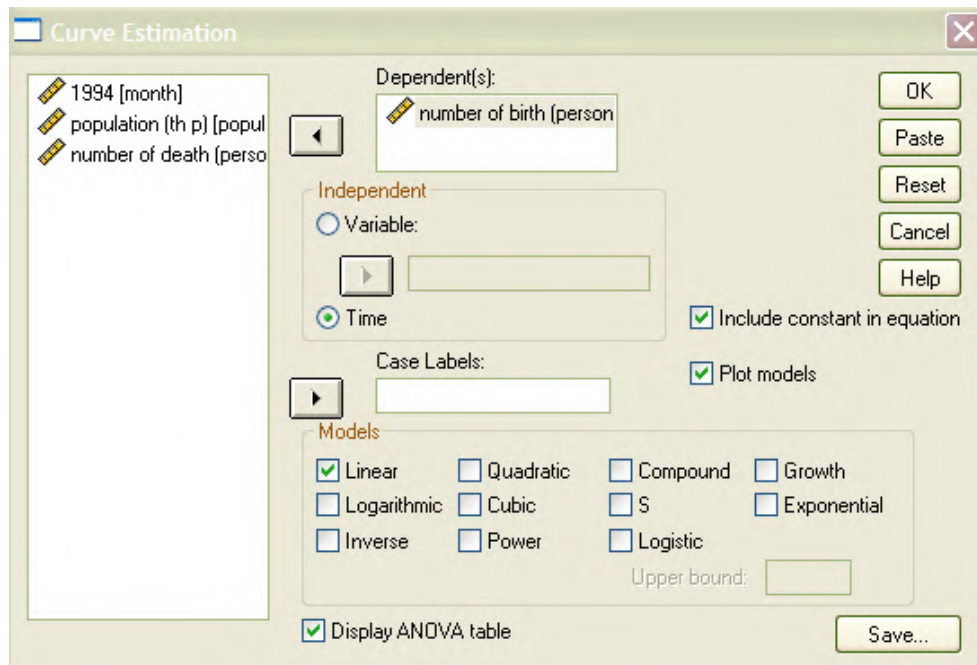


Figure 51 – Curve Estimation for Linear Trend Model

The results are displayed in the Output window.

ANOVA

	Sum of Squares	df	Mean Square	F	Sig.
Regression	536259,1	1	536259,084	3,261	,101
Residual	1644363	10	164436,258		
Total	2180622	11			

According to ANOVA table, the linear regression model is not significant. (Or significant at higher significance level than 10.1%)

Coefficients

	Unstandardized Coefficients		Standardized Coefficients	t	Sig.
	B	Std. Error	Beta		
Case Sequence (Constant)	-61,238	33,910	-,496	-1,806	,101
	10031,212	249,573		40,194	,000

Linear regression line: $\hat{y} = b_0 + b_1 t = 10031.212 - 61.238t$

b_0 : In December 1993, the average number of birth was 10031.212 persons.

b_1 : The monthly average growth rate is -62.238 persons.

c) Give estimation for the number of birth in January 1995!

We can save predicted values in the *Analyze / Regression / Curve Estimation / Save* menu. There are two options for predict cases. The first one is the 'Predict from estimation period through last case', when values are predicted for all cases in the file, based on the cases in the estimation period. The second option ('Predict through') can be used to forecast values beyond the last case in the time series. Now we should predict through 13 observations, because we have 12 data for 1994, and the 13th is January 1995.

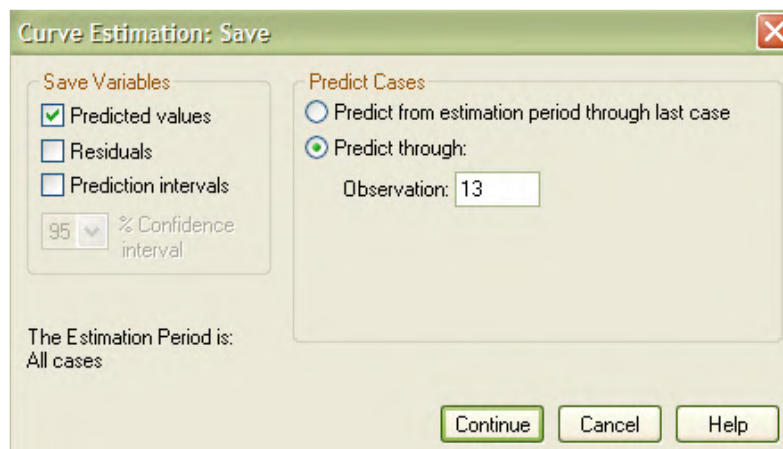


Figure 52 – Curve Estimation: Save Dialog Box

The results are saved in the Data Editor window.

	month	population	birth	death	FIT 1
1	1	10273	10238	13888	9969,97436
2	2	10270	9285	12825	9908,73660
3	3	10267	10105	12516	9847,49883
4	4	10265	9617	11753	9786,26107
5	5	10262	9548	12328	9725,02331
6	6	10260	9717	11839	9663,78555
7	7	10258	9965	11848	9602,54779
8	8	10257	9980	11722	9541,31002
9	9	10256	9844	10968	9480,07226
10	10	10252	9021	12542	9418,83450
11	11	10249	8740	11743	9357,59674
12	12	10246	9538	12917	9296,35897
13	9235,12121

Figure 53 – Predicted Number of Birth

Example 2 The following are data about railway transportation.

Year	Quarters	Railway transport (thousand tons)	Year	Quarters	Railway transport (thousand tons)
1990	1	20 516	1992	3	11 044
1990	2	20 674	1992	4	14 806
1990	3	21 736	1993	1	10 924
1990	4	24 796	1993	2	9 870
1991	1	15 510	1993	3	10 579
1991	2	14 511	1993	4	13 456
1991	3	14 557	1994	1	4 920
1991	4	18 510	1994	2	6 354
1992	1	13 974	1994	3	6 834
1992	2	13 294	1994	4	8 599

Problem

- a) Create a new SPSS data set!

After opening a new SPSS data, we should type the railway transportation figures in the Data View. Then select *Data / Define Dates...* from the menus.



Figure 54 – Define Dates Dialog Box

For this exercise we need cases defined in years and quarters as well. Then select the ‘Years, quarters’ option from the box. The first case is the first quarter in 1990. Then the data set is ready with 3 new columns: Year, Quarter and Date.

b) Create a sequence chart!

A Sequence Chart can be created by clicking on *Analyze / Forecasting / Sequence...*

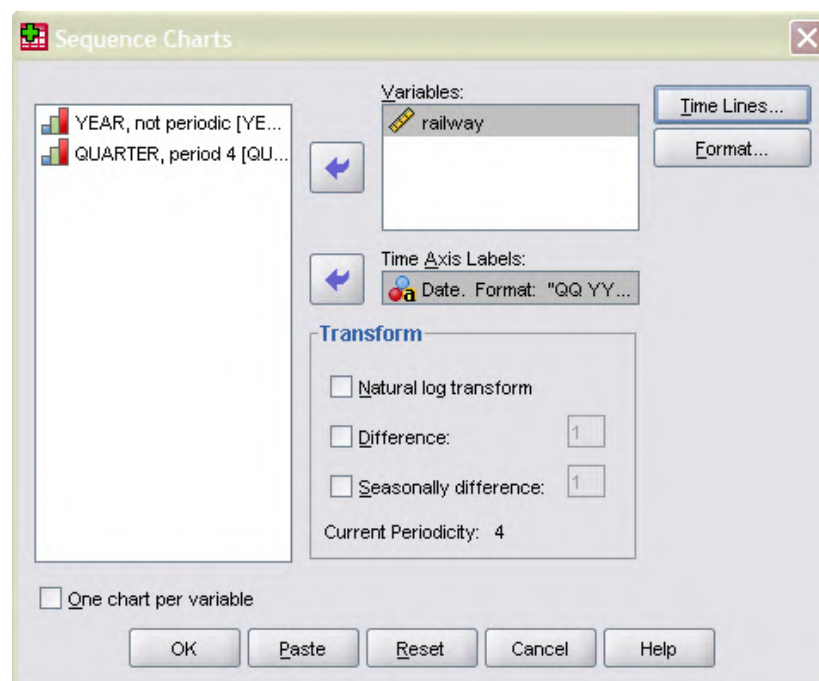


Figure 55 – Sequence Chart Dialog Box

The variable which should be drawn is the transportation, and the time axis is the created date variable. We suppose that we have seasonality in the model, thus add reference lines to the chart by clicking on *Time Lines*.

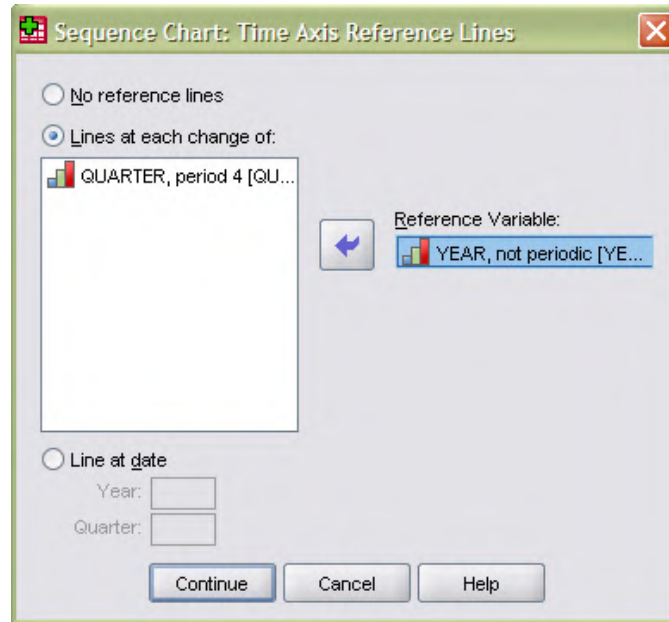


Figure 56 – Time Axis Reference Line Dialog Box

We create lines at each change of quarters. The chart will be displayed in Output window.

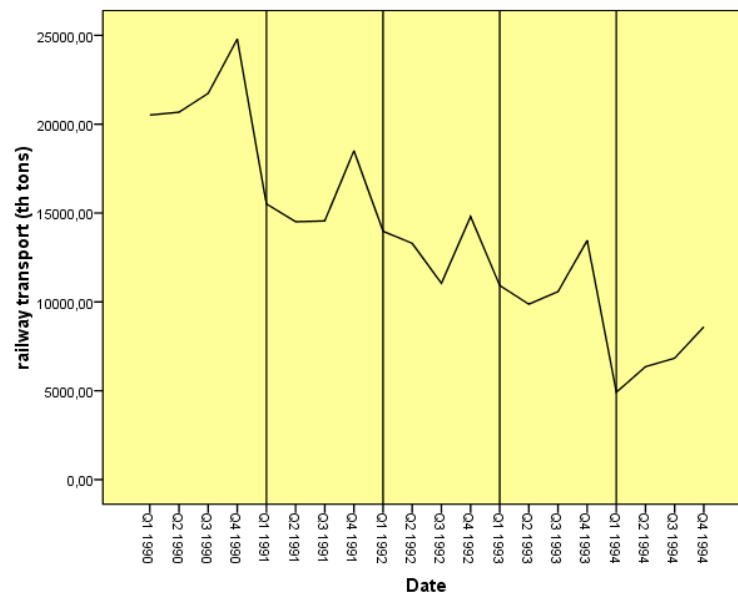


Figure 57 – Time Axis Reference Line Dialog Box

c) Do a seasonal decomposition!

According to the chart edited previously, we realized that there is a seasonal component in the model. The seasonal decomposition model type should be additive, because the amplitude of both the seasonal and irregular variations do not change as the level of the trend rises or falls.

For a seasonal decomposition from the menus choose *Analyze / Forecasting / Seasonal Decomposition / Variable: railway transport*.

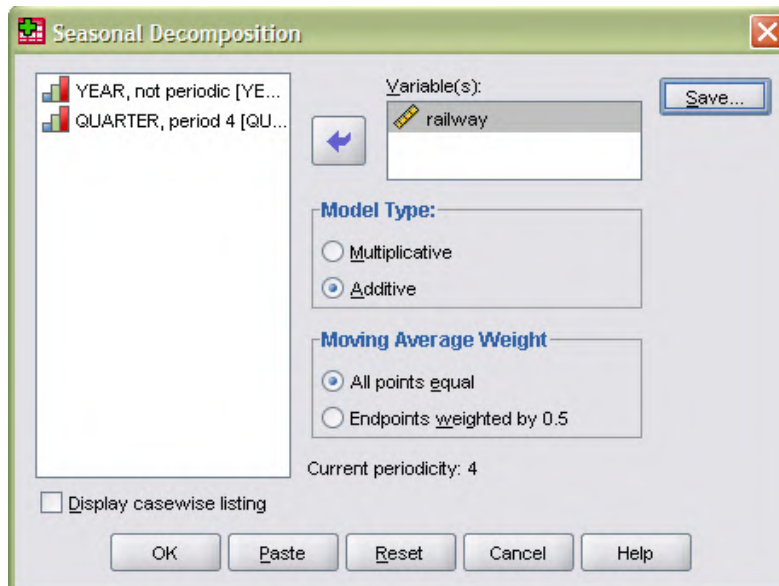


Figure 58 – Seasonal Decomposition Dialog Box

The results will be displayed in the Data Editor window. We will get 4 new columns: ERR_1; SAS_1; SAF_1; STC_1,

where

ERR_1 is the error component.

SAS_1 is the seasonal adjusted series, so the component without seasonality (railway transportation – saf_1).

SAF_1 contains the seasonal factors.

STC_1 is the smoothed component.

For visualization, select *Analyze / Forecasting / Sequence Charts* according to the followings:

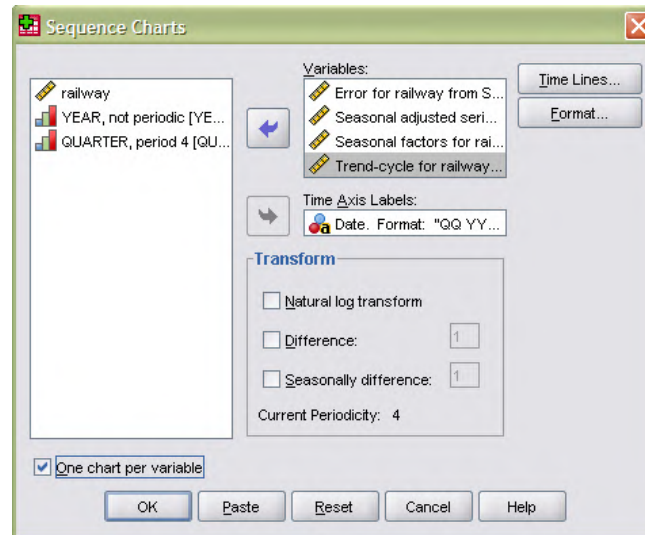


Figure 59 – Seasonal Decomposition Dialog Box

By clicking on ‘One chart per variable’ we will get four different charts.

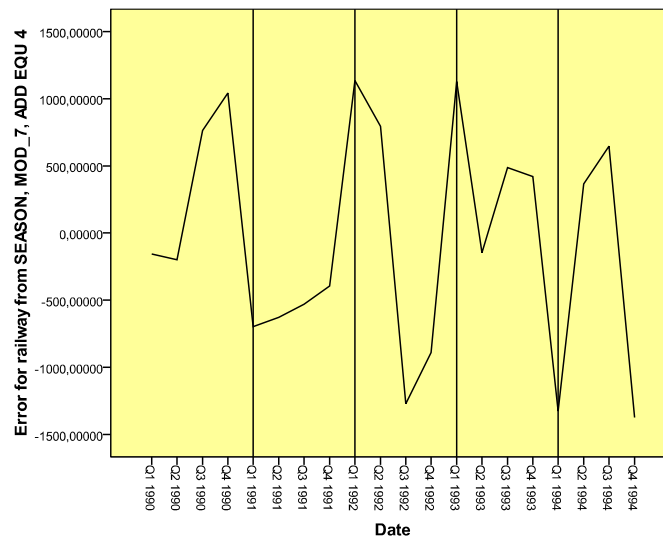


Figure 60 – Error Component

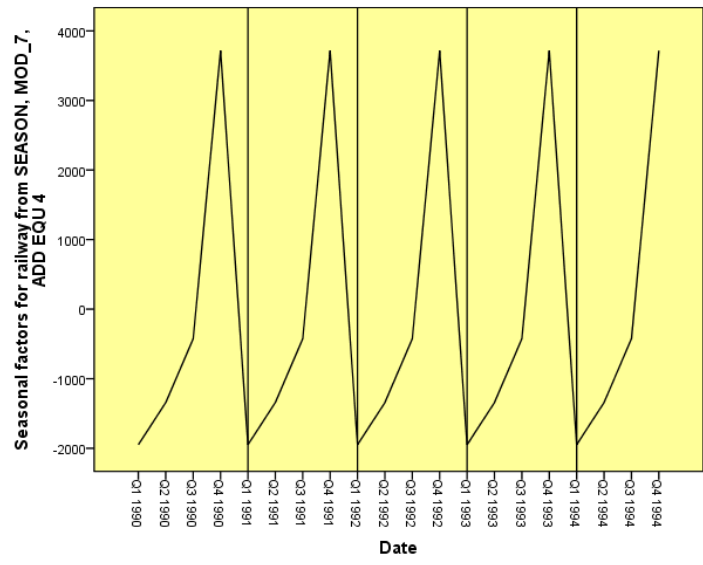


Figure 61 – SAF_1: Seasonal Component

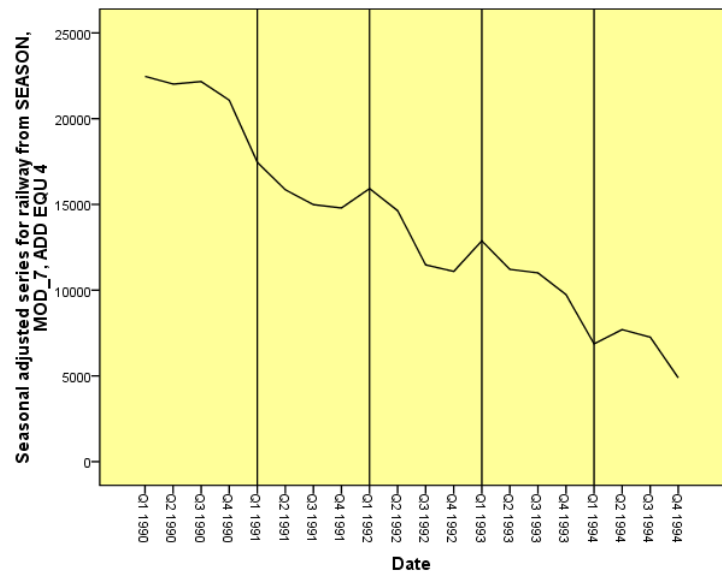


Figure 62 – SAS_1: Component without Seasonality

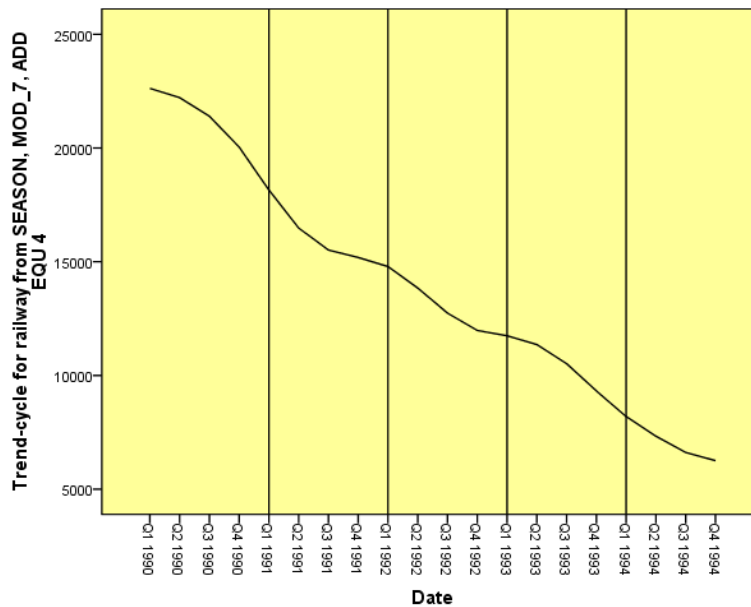


Figure 63 – STC_1: Smoothed Trend-cycle Component

Moreover, we will find Seasonal Factors in the Output window due to the seasonal decomposition. Seasonal factors are used for additive decomposition. The seasonal adjustments are added to the seasonally adjusted series to obtain the observed values.

Seasonal Factors

Series Name:railway

Period	Seasonal Factor
1	-1948.49688
2	-1341.87188
3	-426.13438
4	3716.50313

The first three factors mean that the trend has made too high an estimate of the quantity of railway transport. The fourth seasonal factor is a positive figure, because the trend has underestimated the real value of the railway transport.

d) Give estimation for the railway transport in Quarter III 1995!

Select: *Analyze / Regression / Curve Estimation / Save...*

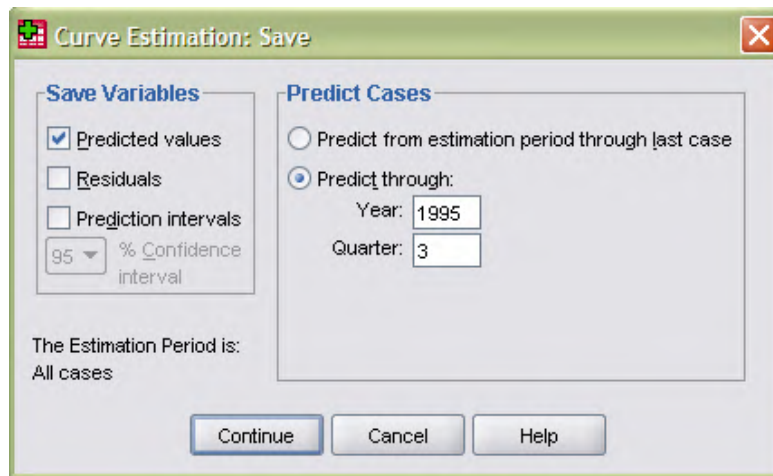


Figure 64 – Forecasted Values

We should just define the end of prediction period, which is the Quarter 3 in the year of 1995. Then we will get the results.

	railway	YEAR_	QUARTER_	DATE_	ERR_1	SAS_1	SAF_1	STC_1	FIT_2
1	20516,00	1990	1	Q1 1990	-156,74722	22464,49688	-1948,49688	22621,24410	21482,45714
2	20674,00	1990	2	Q2 1990	-198,29583	22015,87188	-1341,87188	22214,16771	20670,95639
3	21736,00	1990	3	Q3 1990	762,11944	22162,13438	-426,13438	21400,01493	19859,45564
4	24796,00	1990	4	Q4 1990	1040,77500	21079,49688	3716,50313	20038,72188	19047,95489
5	15510,00	1991	1	Q1 1991	-695,44722	17458,49688	-1948,49688	18153,94410	18236,45414
6	14511,00	1991	2	Q2 1991	-626,55833	15852,87188	-1341,87188	16479,43021	17424,95338
7	14557,00	1991	3	Q3 1991	-530,54722	14983,13438	-426,13438	15513,68160	16613,45263
8	18510,00	1991	4	Q4 1991	-393,22500	14793,49688	3716,50313	15186,72188	15801,95188
9	13974,00	1992	1	Q1 1992	1135,88611	15922,49688	-1948,49688	14786,61076	14990,45113
10	13294,00	1992	2	Q2 1992	794,10833	14635,87188	-1341,87188	13841,76354	14178,95038
11	11044,00	1992	3	Q3 1992	-1269,43611	11470,13438	-426,13438	12739,57049	13367,44962
12	14806,00	1992	4	Q4 1992	-888,44722	11089,49688	3716,50313	11977,94410	12555,94887
13	10924,00	1993	1	Q1 1993	1128,55278	12872,49688	-1948,49688	11743,94410	11744,44812
14	9870,00	1993	2	Q2 1993	-145,89167	11211,87188	-1341,87188	11357,76354	10932,94737
15	10579,00	1993	3	Q3 1993	487,45278	11005,13438	-426,13438	10517,68160	10121,44662
16	13456,00	1993	4	Q4 1993	420,21944	9739,49688	3716,50313	9319,27743	9309,94586
17	4920,00	1994	1	Q1 1994	-1325,00278	6868,49688	-1948,49688	8193,49965	8498,44511
18	6354,00	1994	2	Q2 1994	366,21944	7695,87188	-1341,87188	7329,65243	7686,94436
19	6834,00	1994	3	Q3 1994	647,30000	7260,13438	-426,13438	6612,83437	6875,44361
20	8599,00	1994	4	Q4 1994	-1371,92847	4882,49688	3716,50313	6254,42535	6063,94286
21	.	1995	1	Q1 1995	5252,44211
22	.	1995	2	Q2 1995	4440,94135
23	.	1995	3	Q3 1995	3629,44060

Figure 65 – Predicted Railway Transport

Note that we had seasonal component, thus we should modify the result:

$$y_{Q3, 1995} = 3629.44 - 426.134 = \mathbf{3203.306}$$

According to the linear trend model, 3203.306 tons will be the railway transportation in Quarter 3, 1995.

II. EXERCISES FOR SPSS

Exercise 1

The data below are about students at the University of Miskolc.

Name	Age	Place of residence	School year	Grade (Statistics)
T.E.	24	Miskolc	3	3
B.N.	32	Miskolc	2	4
H.L.	22	Felsőzsolca	3	2
K.O.	35	Miskolc	3	4
V.I.	27	Mályi	2	5

Create an SPSS data set!

Exercise 2

The data below are about the result of the final competition of 'Who knows more about Statistics?'

	Gender	Age	Award	Prize (HUF)
1	Woman	22	1	55 000
2	Woman	23	2	30 000
3	Man	22	3	15 000
4	Man	21	4	0
5	Woman	22	5	0
6	Woman	20	6	0
7	Woman	21	7	0
8	Man	23	8	0

Create an SPSS data set!

Exercise 3

	2000	2004	2005
University student	47498	66309	70153
College student	54889	88384	109412
Number of the teachers	17302	19103	18098
Number of institutions	77	91	90
Number of students in the % of the 18-22-year-old population	10.4	13.9	15

Source: Ketskeméthy – Izsó (2006)

- Create an SPSS data set!
- Determine the number of 18-22-year-old population in 2000, 2004 and 2005!

Exercise 4

Open the *Employee_data.sav* file!

Topic 1: Transform / Select Data

- a) What is the proportion of custodials?
- b) What is the proportion of women within managers?

Topic 2: Graphs

- a) Create a column diagram about the proportion of employees grouped by *gender*! Embellish the graph! Put the value of proportions into the chart!
- b) Transform this column diagram into a pie chart!
- c) Create a scatter plot about *month since hire* and *beginning salary* if you set markers by *gender*! Embellish the graph!
- d) Create a scatter plot about *month since hire* and *previous experience* if you set markers by *employment category*! Embellish the graph!
- e) Define simple box plot about *previous experience*! Embellish the graph!
- f) Define simple box plot about the *month since hire* categorized by the *employment category*! Embellish the graph!
- g) Define box plot about the *previous experience* categorized by the *employment category* clustered by *gender*! Embellish the graph!
- h) Create a graph to test the normal distribution of *beginning salary*!

Topic 3: Central Tendencies, Measures of Distribution, Measures of Asymmetry

- a) Define the central tendencies of *month since hire*!
- b) Define the characteristics of distribution of *previous experience*!
- c) What is the average salary of employees belonging to the minority?

Topic 4: Estimation and Hypothesis Testing

- a) Define a 95% confidence interval for the *previous experience*!
- b) Define a 90% confidence interval for the *beginning salary*!
- c) Define a 98% confidence interval for *month since hire*!
- d) Test the hypothesis that the *beginning salary* of the employees equals \$35 000. ($\alpha = 5\%$)
- e) Test the hypothesis that the *beginning salary* of the employees equals \$15 000. ($\alpha = 10\%$)
- f) Test the hypothesis that the *previous experience* of employees equals 95 month. ($\alpha = 10\%$)
- g) Test the hypothesis that the *previous experience* of the managers equals 100 month. ($\alpha = 10\%$)
- h) Test the hypothesis that the *current salary* of the custodial and manager are equal! ($\alpha = 10\%$)
- i) Test the hypothesis that the *current salary* of the employees belonging and not belonging to the minority are equal! ($\alpha = 5\%$)
- j) Are the *beginning salary* of men and women equal? ($\alpha = 5\%$)
- k) Test the normality of *beginning salary*! ($\alpha = 5\%$)
- l) Test the hypothesis that the *previous experience* follows normal distribution! ($\alpha = 5\%$)

Topic 5: Statistical Dependence

- a) Create a crosstabs about *gender* and *minority classification*!
- b) Is there any relationship between *gender* and *minority classification*?
- c) Is there any relationship between *employment category* and *minority classification*?
- d) Is there any relationship between *minority classification* and *current salary*?
- e) Is there any relationship between *minority classification* and *previous experience*?
- f) Is there any relationship between *gender* and *previous experience*?

Topic 6: Correlation and Linear Regression

- a) Is there any relation between *previous experience* and *month since hire*?
- b) Determine a linear relation between the *month since hire* and *previous experience* of employees!
- c) Define a 90% confidence interval for its b_0 and b_1 parameters!
- d) Define a 90% confidence interval for the y variable!

Topic 7: Multiple Correlation and Linear Regression

- a) Is there any relation between *previous experience*, *gender*, *age* and *month since hire*?
- b) Determine a linear relation between the *month since hire* (y) and *previous experience* (x_1), *gender* (x_2), and *age* (x_3) of employees!

Topic 8: Curvilinear Regression

- a) Which regression model fit the most to the relation between *month since hire* and *previous experience*?
- b) Which regression model fit the most to the relation between *current salary* and *previous experience*?

Exercise 5

Open the Cars.sav file!

Topic 1: Transform / Select Data

- a) How old are the cars? Create a new variable as age!
- b) What is the ratio of American, European and Japanese cars within cars with higher consumption than 20 miles per gallon?
- c) What is the ratio of those American cars which have 4-6-8 cylinders?

Topic 2: Graphs

- a) Create a column diagram about the proportion of cars grouped by the *county of origin*! Embellish the graph! Put the value of proportions into the chart!
- b) Create pie chart about the proportion of cars grouped by *country of origin*! Embellish the graph!
- c) Create a scatter plot of *horsepower* and *vehicle weight* if you set markers by the *country of origin*! Embellish the graph!
- d) Define simple box plot of *horsepower*! Embellish the graph!

- e) Define simple box plot of *time to accelerate* categorized by the *country of origin*! Embellish the graph!
- f) Define simple box plot of petrol consumption (*miles per gallon*) categorized by the *country of origin*! Embellish the graph!
- g) Define horizontal box plot of *horsepower* categorized by the number of *cylinders* clustered by the *country of origin*! Embellish the graph!
- h) Create a graph to check the normality of *horsepower*!

Topic 3: Central Tendencies, Measures of Distribution, Measures of Asymmetry

- a) Define the central tendencies of *horsepower*!
- b) Define the measures of distribution of *horsepower*!
- c) Define the measures of asymmetry of *horsepower*!
- d) Define the quartiles of *horsepower*!
- e) Define the characteristics of distribution of *time to accelerate*!
- f) Define the characteristics of distribution of *vehicle weight*!
- g) Define the characteristics of distribution of *engine displacement*!
- h) Define the characteristics of distribution of consumption (*miles per gallon*)!
- i) What is the average consumption of those cars which engine displacement is higher than 100 cubic inch?

Topic 4: Estimation and Hypothesis Testing

- a) Define a 95% confidence interval for the *vehicle weight*!
- b) Define a 90% confidence interval for the *horsepower*!
- c) Define a 98% confidence interval for the *time to accelerate*!
- d) Test the hypothesis that the average consumption of cars is 20 *miles per gallon*! ($\alpha = 5\%$)
- e) Use One Sample T Test to determine whether or not the average *miles per gallon* significantly differ from 24 at 10% significance level!
- f) Test the hypothesis that the average *horsepower* of cars is 100! ($\alpha = 5\%$)
- g) Test the hypothesis that the average consumption of Japanese and American cars is the same! ($\alpha = 5\%$)
- h) Test the hypothesis that the average consumption of European and American cars is the same! ($\alpha = 10\%$)
- i) Check if the *horsepower* follows a normal distribution or not!

Topic 5: Statistical Dependence

- a) Create a crosstabs from the *model year* and the *country of origin*!
- b) Create a crosstabs from the *number of cylinders* and the *country of origin*!
- c) Is there any relationship between the *country of origin* and *engine displacement*?
- d) Is there any relationship between the *country of origin* and *horsepower*?
- e) Is there any relationship between the *country of origin* and *vehicle weight*?

Topic 6: Correlation and Linear Regression

- a) Is there any relationship between the *horsepower* and consumption (*miles per gallon*)?
- b) Is there any relationship between the *engine displacement* and consumption (*miles per gallon*)?

- c) Is there any relationship between the *vehicle weight* and consumption (*miles per gallon*)?
- d) Determine a linear relationship between the consumption (*miles per gallon*) and *horsepower*!
 - 1. Estimate the average consumption! ($\pi=95\%$)
 - 2. Estimate the b_0 , b_1 parameters! ($\pi=95\%$)
- e) Determine a linear relationship between the *time to accelerate* and *horsepower*!
 - 3. Estimate the b_0 , b_1 parameters! ($\pi=90\%$)
 - 4. Estimate the average consumption! ($\pi=90\%$)

Topic 7: Multiple Correlation and Linear Regression

- a) Is there any relationship between the *horsepower*, consumption (*miles per gallon*), *engine displacement*, and *vehicle weight*?
- b) Determine a linear relationship between the consumption (y) and *horsepower* (x_1), *engine displacement* (x_2), and *vehicle weight* (x_3)!

Topic 8: Curvilinear Regression

- a) Determine the relationship between the consumption (*miles per gallon*) and *horsepower*! Which regression model fit the most?
- b) Determine the relationship between the *time to accelerate* and the *vehicle weight*! Which regression model fit the most?

Exercise 6

The data below are about Ultrasound equipments:

Sequence Number	Running time (year)	Repair time (hour)
1	6	42
2	2	22
3	9	75
4	12	96
5	5	32
6	3	25
7	4	33
8	6	38
9	11	91
10	8	61
11	1	17
12	7	51
13	4	31
14	3	35
15	5	47
16	9	65
17	2	24
18	6	44
19	3	29
20	11	88

Problem:

- Create an SPSS data set!
- Analyze the characteristics of distribution (mean, standard deviation, mode, median, quartiles, minimum, maximum values and asymmetry) of the running time!
- Draw a (horizontal) box plot about repair time & analyze this! Embellish the graph! Insert a title, name and measurement of the 'x' axis!
- Is there any dependence between the running time and repair hour? Determine C, r and r^2 measures!
- Define the parameters of linear regression!
- Test the regression and the parameters at 5% significance level!
- Estimate the b_1 parameter! ($\pi=98\%$)

Exercise 7

The following table shows the age (X) and blood pressure (Y) of 12 women.

Age	56	42	72	36	63	47	55	49	38	42	65	37
Blood pressure (Hg mm)	147	125	160	118	149	128	150	145	115	140	165	110

Problem:

- Create a new SPSS data set!
- Which regression model fit the most? Determine the relation between age and blood pressure!
- Create a graph about their relation! Attach the regression equation to the chart! Embellish the graph!
- Determine the strength of relationship!

Exercise 8

After a Statistics exam we made a research about the time spent on learning for the exam and the result. The following are the results of a random sampling.

Name	Time spent on learning for the exam (hour)	Result of exam (points)
B. M.	2	9
B. E.	3	15
F. O.	5	20
H. B.	10	50
I. L.	15	58
N. N.	20	75
O. A.	22	80
P. L.	18	65
V. ZS.	25	100
ZS. B.	30	95

Problem:

- Create a new SPSS data set!
- Determine the strength and nature of this relationship!

Exercise 9

Ten students were ranked according to their mathematical and musical ability:

Student Ability	A	B	C	D	E	F	G	H	I	J
Mathematics	1	2	3	4	5	6	7	8	9	10
Music	3	4	1	2	5	7	10	6	8	9

Problem:

- Create a new SPSS data set!
- Determine the strength of the dependence between the abilities!

Exercise 10

The following are data about the electricity consumption of a village in thousand kW per hours.

Year	Quarter			
	I	II	III	IV
1991	344	305	287	332
1992	502	440	401	474
1993	704	619	565	633
1994	1 009	880	787	915
1995	1 467	1 271	1 085	1 192

Problem:

- Create an SPSS data set!
- Define the linear trend model for the electricity consumption!
- Create graph about the trend! Attach the equation to the line!
- Test the normality of electricity consumption using the hypothesis testing!
- Test the normality of electricity consumption using histogram!
- Define the seasonal components!
- Determine the electricity consumption for Quarter II in 1996!

Exercise 11

The relation between juice consumption and daily maximum temperature was examined through 20 days on the beach. The following are data about the results.

Day	Juice consumption (liter)	Daily maximum temperature (°C)
1	520	25
2	534	26
3	610	28
4	780	32
5	708	27
6	639	25
7	486	23
8	423	20
9	452	22
10	597	29

11	640	30
12	657	31
13	678	30
14	620	27
15	635	28
16	610	26
17	585	25
18	627	27
19	608	26
20	720	30

Problem:

- Create a new SPSS data set!
- Determine a linear relation between juice consumption and temperature!
- Create a graph about their relation! Attach the regression equation to the chart!
Embellish the graph!
- Determine the strength of relationship!

Exercise 12

In a factory the production of a machine and the amount of faulty product were examined. The data below are about the result of 20 days.

Day	Production (1000 bottles per hour)	Faulty products (1000 bottles per day)
1	20	9.0
2	18	9.0
3	26	11.4
4	21	9.5
5	22	9.5
6	30	38.5
7	23	9.7
8	29	25.5
9	30	38.0
10	22	10.0
11	19	9.0
12	31	56.0
13	17	9.0
14	24	10.5
15	25	10.4
16	27	14.5
17	25	11.0
18	27	13.5
19	24	10.0
20	17	8.5

Problem:

- Create a new SPSS data set!

- b) Which regression model fit the most? Determine the parameters of regression!
- c) Create a graph about their relation! Attach the regression equation to the chart!
Embellish the graph!
- d) Determine the strength of relationship!

Exercise 13

The following are data about the change in CPI (previous year = 100%) and the unemployment rate in Germany between 1972 and 1997.

Year	Germany	
	Consumer Price Index (%)	Unemployment rate (%)
1972	5.5	1.1
1973	6.9	1.2
1974	7.0	2.6
1975	6.0	4.7
1976	4.5	4.6
1977	3.7	4.5
1978	2.7	4.3
1979	4.1	3.8
1980	5.6	3.8
1981	6.3	5.5
1982	5.3	7.5
1983	3.3	9.1
1984	2.4	9.0
1985	2.2	9.5
1986	0.1	9.0
1987	0.2	7.6
1988	1.3	7.6
1989	2.8	6.9
1990	2.7	6.2
1991	3.5	6.6
1992	4.0	7.7
1993	4.5	8.8
1994	2.7	9.6
1995	1.8	9.3
1996	1.5	11.5
1997	1.8	9.8

Problem

- a) Create a new SPSS data set!
- b) Which regression model fit the most? Determine the parameters of regression!
- c) Test the regression and the parameters at 5% significance level!
- d) Create a graph about their relation! Attach the regression equation to the chart!
Embellish the graph!
- e) Determine the strength of relationship!
- f) Determine a 90% confidence interval for the b_0 parameter!

Exercise 14

Create a new data file according to this questionnaire!

1. How often do you go to cinema?		2	3	1	2	4	2	3	3	4	3	4	5
1	More times a week												
2	Once a week												
3	Every second week												
4	Once a month												
5	Rarely												
6	Never												
2. Gender		1	1	1	1	1	1	2	2	2	2	2	2
1	Male												
2	Female												
3. Date of birth: ... (year/month/day)													
1	1985/04/12	7											
2	1981/12/30	8											
3	1991/11/11	9											
4	1992/08/03	10											
5	1985/05/14	11											
6	1990/07/01	12											

Problem:

- g) How old are the people? Create a new variable as age!
- h) Create a bar chart about the result of the first question (%)! Embellish the chart!
- i) Create a pie chart about gender! Use the 3D design! Embellish the chart!
- j) Create a contingency table about the gender and the time how often people go to cinema!
- k) Is there any dependence between the gender and the time how often people go to cinema?
- l) Is there any dependence between the age and the time how often people go to cinema?

Exercise 15

The influence factors of the salary were examined at a company. The following are data about 45 employees.

Salary (HUF/hour)	Years since hire	Age	Gender	Qual.	Salary (HUF/hour)	Years since hire	Age	Gender	Qual.
188	25	45	1	1	171	9	36	1	1
157	16	45	0	0	142	7	26	1	0
165	30	51	0	0	150	10	26	0	0
124	5	39	0	0	156	15	28	0	0
139	12	31	0	0	154	20	41	0	0
165	17	34	0	1	176	25	43	1	1
158	10	31	0	1	137	13	42	0	0
224	24	44	1	1	130	7	23	0	0
169	17	45	1	1	155	7	44	0	1

114	6	25	1	0	234	33	52	1	1
160	11	48	0	0	200	25	42	1	1
154	27	46	1	0	228	24	44	1	1
150	14	30	1	0	161	16	33	0	1
130	7	23	1	0	148	5	43	0	1
198	31	56	1	1	127	2	20	0	1
159	16	33	0	1	195	22	39	1	1
154	16	32	0	0	237	27	40	1	1
174	17	35	1	0	163	21	46	0	0
126	7	44	0	0	201	18	41	1	1
162	12	29	0	1	137	5	23	1	1
181	26	46	1	1	233	27	45	1	1
146	10	47	0	0	180	15	42	1	1
152	7	30	1	1					

Gender: 0 – man; 1– woman

Qualification: 0 – without qualification; 1 – with qualification

Problem:

- Create a new SPSS data set!
- Create a bar chart about the proportion of gender! Embellish the chart!
- Transform it into a pie chart! Use the 3D design!
- What is the average age of women?
- Is there any dependence between salary and gender?
- Determine an optimal regression for salary!

Exercise 16

The following are data about cars for sale:

Ordinal number	Year of origin	Cylinder	Garage	Colour	Price (Unit)
1	1993	75	0	1	50 000
2	1993	125	0	1	70 000
3	1993	75	0	1	60 000
4	1994	250	1	1	80 000
5	1994	75	0	1	70 000
6	1994	125	1	1	80 000
7	1995	75	0	1	60 000
8	1995	125	0	1	80 000
9	1995	250	0	2	100 000
10	1996	250	1	3	170 000
11	1996	250	1	3	168 000
12	1997	75	1	2	100 000
13	1997	125	1	2	120 000
14	1998	250	0	3	156 000
15	2004	250	1	5	560 000
16	1999	500	1	5	380 000

17	2000	500	1	5	425 000
18	2001	250	0	4	320 000
19	2002	125	1	4	300 000
20	2003	75	1	4	220 000

Codes:

Garage: 0 – The car was not kept in a garage.
1 – The car was kept in a garage.

Colour: 1 - red,
2 - green,
3 - yellow,
4 - blue,
5 - black

Problem:

- Create a new SPSS data set!
- Create a bar chart about the proportion of the colour of cars! Embellish the chart!
- Transform it into a pie chart! Use the 3D design! Embellish the chart!
- What is the average price of yellow cars?
- Determine the central tendencies, the measures of shape, the measures of asymmetry and the quartiles of the price!
- Define a box plot of prices! Embellish the graph!
- Is there any dependence between the price and the model year?
- Is there any dependence between the price and the colour?
- Create a crosstabs about the colour and the storage (garage) of cars!
- Use the K-S Test to test whether the price follows a normal distribution or not!
- Determine a 90% confidence interval for the average prices!

Exercise 17

The following are monthly data about the number of tourist arrived to Wonderful Country between 2009 and 2010:

2008	Number of tourists	2009	Number of tourists	2010	Number of tourists
January	382	January	534	January	997
February	446	February	676	February	748
March	608	March	1012	March	1205
April	692	April	1391	April	1389
May	1029	May	1538	May	1631
June	1174	June	1848	June	1947
July	2219	July	2932	July	2918
August	3080	August	3338	August	4115
September	1597	September	2073	September	2348
October	1241	October	1730	October	1721
November	942	November	1724	November	1357
December	818	December	1669	December	1448

Problem:

- a) Create an SPSS data set!
- b) Define a linear trend model for the number of tourists!
- c) Create graph about the trend!
- d) Test the normality of the number of tourists!
- e) Define the seasonal components!
- f) Make a prediction to May 2011!
- g) Make a prediction to August 2011!

Exercise 18

The following are data about the turnover of a company:

Year	Quarter	Turnover (Million HUF)
1994	I	15
	II	24
	III	26
	IV	39
1995	I	20
	II	28
	III	30
	IV	48
1996	I	26
	II	32
	III	33
	IV	55

Problem:

- a) Create an SPSS data set!
- b) Define the change in turnover by a linear trend model!
- c) Create graph about the trend!
- d) Test the normality of turnover!
- e) Define the seasonal component in Quarter 2!
- f) Make a prediction to Quarter 2 in 1998!

Exercise 19

The following are data about sales (million HUF) of a department store (1998-2001).

Year / Quarter	I	II	III	IV
1998	60	80	100	160
1999	70	85	95	170
2000	80	100	105	165
2001	90	105	110	185

- a) Create a new SPSS data set!
- b) Analyze parameters of a linear trend!
- c) Determine the seasonal factors of the first quarter!
- d) Determine the sales in Quarter II., 2003!

Exercise 20

The data below are about the number of tourists in Hungary between 1988 and 1994.

Year	Quarters	Number of tourists (thousand persons)	Year	Quarters	Number of tourists (thousand persons)
1988	1	687.5	1990	4	1061.2
1988	2	944.7	1991	1	839
1988	3	1212.8	1991	2	1446
1988	4	999.4	1991	3	2274.7
1989	1	839.8	1991	4	1281.5
1989	2	1126.6	1992	1	868.1
1989	3	1423.4	1992	2	1374
1989	4	1164.8	1992	3	1823.9
1990	1	896.2	1992	4	1319.3
1990	2	1307.8	1993	1	854
1990	3	1887.8			

- Is there any trend in this model? (Normality test)
- Create a graph from the time series!
- Which seasonal decomposition should you use? Why?
- Do a seasonal decomposition! Analyze the parameters and the seasonal factors!
- Create graphs from the seasonal factors (saf_1, sas_1, stc_1)!
- Determine the number of tourists for the 2nd, 3rd and 4th quarter of 1993!

REFERENCES

- [1] Aczel, A.: *Complete Business Statistics*. New York: Richard Irwin, 1996
- [2] Brooks, C.: *Introductory Econometrics for Finance*. Cambridge; Second Edition: Chapter 2., 2008
- [3] Defusco, R. A. –McLeavey, D. W. –Pinto, J. E. –Runkley, D.E.: *Quantitative Investment Analysis*, CFA Series; Second Edition: Chapter 6, 7, 8, 10., 2007
- [4] Domán Cs. – Szilágyi R. – Varga B.: *Statisztikai elemzések alapjai II*. Miskolci Egyetem, Miskolc, 2009
- [5] Ketskeméthy L. – Izsó L.: *Bevezetés az SPSS programrendszerbe* ELTE Eötvös. Kiadó, Budapest, 2005
- [6] SPSS 16.0 Tutorial
- [7] SPSS 17.0 Tutorial
- [8] SPSS 19.0 Tutorial
- [9] SPSS Online Training Workshop, Central Michigan University: <http://www.cst.cmich.edu/users/lee1c/spss/index.htm> (accessed: 06-01-2011)
- [10] Statistics Online Tutorial: stattrek.com (accessed: 05-01-2011)

Csernely_data.sav file can be downloaded from the website of the University of Miskolc / Faculty of Economics (www.gtk.uni-miskolc.hu) / Institute of Business Information and Methods / Department of Business Statistics and Economic Forecasting: <http://193.6.12.228/uigtk/uise/index.html>