STAR USER MANUAL

Beta Version 2012

Crop Research Informatics Laboratory INTERNATIONAL RICE RESEARCH INSTITUTE

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1. Introduction

Statistical Tool for Agricultural Research (STAR) is a computer program for data management and basic statistical analysis of experimental data. It was created using the Eclipse Rich Client Platform (RCP) and uses the R language and environment for statistical computing and graphics. It has a user-friendly graphical interface where items are accessible via pull-down menus.

STAR has been developed primarily for the analysis of data from agricultural field trials, but many of the features can be used for analysis of data from other sources.

The main modules and facilities are

- 1. Data management with a spreadsheet
- 2. Descriptive statistics
- 3. Analysis of variance for basic experimental designs
- 4. Correlation and linear regression analysis
- 5. Graphics
- 6. Utilities for randomization and electronic field book
- 7. Non-parametric data analysis

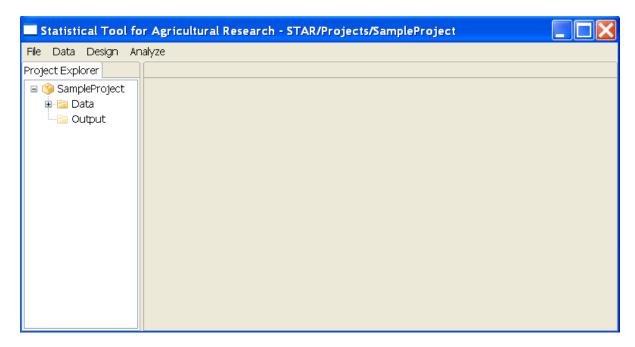
The STAR user's manual provides step-by-step instructions on how to perform certain tasks of interest to users. Screen images have been included as deemed helpful.

Menu items, names of dialogs and form controls are in **bold** letters. Filenames, variable names and directories are *italicized*. Menus appear in the form **File | Open**, which means "choose **Open** from the **File** menu". All of the examples used in this manual are included when the STAR package is installed.

2. Overview of STAR

2.1. Starting STAR

Upon launching the STAR, the main window will appear.



2.2. STAR Menu Bar

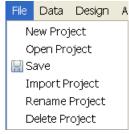
Most of the tasks that you perform in STAR start with menu selections. STAR has a menu bar (Figure 3.2 - 1) and five tab pages (Figure 3.2 - 2). The same main menu is displayed in all tab pages.



There are four pull down menus available in STAR and this are:

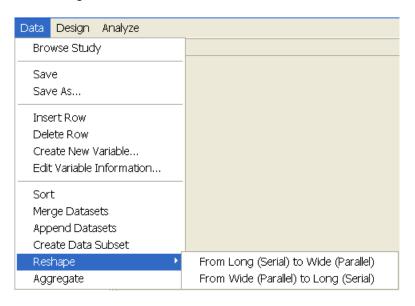
File Menu

contains creating new project, open existing project, saving, importing project, renaming project and deleting project. It contains the following submenu items:



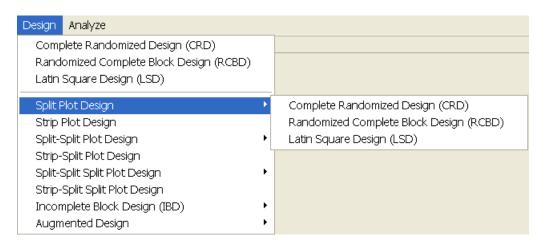
Data Menu

contain items for reading and manipulating datasets. It contains the following submenu items:



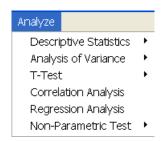
Design Menu

contain items for performing randomization. It contains the following submenu items:



Analyze Menu

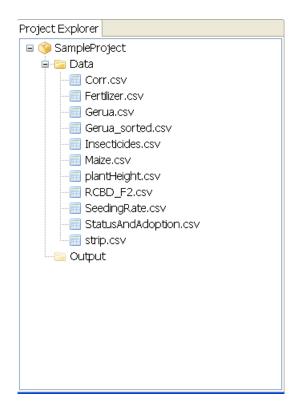
contain items to perform statistical analysis. It contains the following submenu items:



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2.3. Project Explorer

Project Explorer can be seen at the right-hand side of the main window. For first-time users, a default project named *SampleProject* is available with *Data* folder and *Output* folder inside it. The *Data* folder contains sample datasets that will be used in this tutorial. The *Output* folder, on the other hand, will contain all output files that will be created when an analysis is performed.



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3. Data Manipulation

3.1. Insert Row(s)

- On the Project Explorer, locate the *data1.csv* file from the *Data* folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.
- Select any row(s) or any cell in the row(s) where the user wants to insert new row(s) above it. To select several cells/rows, click any cell/row, then hold the Ctrl key and click on another cell/row. The number of selected rows is the number of row that will be inserted.
- To insert row, click Data | Insert Row from the main window of STAR or click on the
 Insert Row icon in the Data Viewer tool bar.
- If there is no row selected before executing the Insert Row function, the user will be prompted if the user wants to insert a row after the last row of the data.

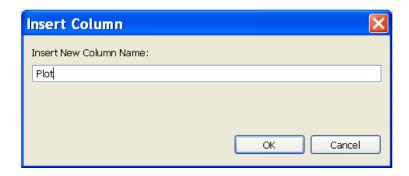
3.2. Delete Row(s)

- On the Project Explorer, locate the *data1.csv* file from the *Data* folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.
- Select any row(s) or any cell in the row(s) that the user wants to delete. To select several cells/rows, click any cell/row, then hold the Ctrl key and click on another cell/row.
 The number of selected rows is the number of row that will be deleted.
- To delete row(s), click Data | Delete Row from the main window of STAR or click on the
 Delete Row icon in the Data Viewer tool bar.

3.3. Inserting Column(s)

- On the Project Explorer, locate the *data1.csv* file from the *Data* folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.
- Select a column where the user wants a new column to be inserted before it by clicking on the column name. A check icon will appear on the column header.
- To insert column(s), click **Data** | **Insert Column** from the main window of STAR or click on the Insert Column icon in the Data Viewer tool bar.

The user will be ask to specify the name of the new column, say, *Plot*.



Click OK.

3.4. Deleting Column(s)

- On the Project Explorer, locate the data1.csv file from the Data folder of the project named SampleProject. Double-click the file to view it in the Data Viewer.
- Select the column or one of the columns to be deleted by clicking on the column name. A check icon will appear on the column header.
- To delete column(s), click Data | Delete Column or click on the Insert Column icon in the Data Viewer tool bar.
- Click OK.

3.5. Create New Variables

- On the Project Explorer, locate the *data1.csv* file from the *Data* folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.
- To create new variables, click **Data** | **Create New Variable** from the main window of STAR or click on the Create New Variable icon in the Data Viewer tool bar.
- Specify the required fields and appropriate options.

Name

This is the name of the new variable to be created.

Values

There are two options to determine the values of the new variable. The user can either transform any existing numeric variables or concatenate the values of any variables in the data set.

If transformation is selected, the user needs to specify the following:

Data Manipulation 6 | Page

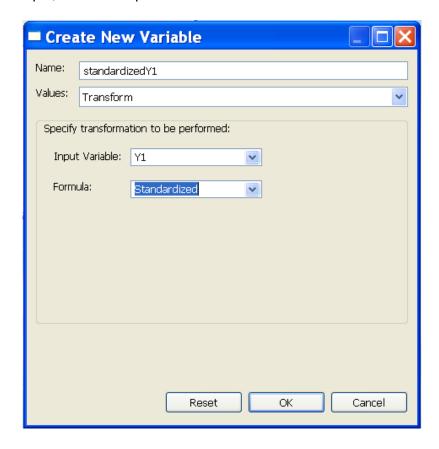
Input Variable

This is variable where the transformation will be performed.

Formula

Transformation options available are logarithm, natural logarithm, square root, power, exponential and standardized.

For the example, select the options as shown below:



If concatenation of columns is selected, the user needs to specify the columns to concatenate.

Click OK.

3.6. Edit Variable Information

- On the Project Explorer, locate the *data1.csv* file from the *Data* folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.
- To edit variable information, click **Data | Edit Variable Information** from the main window of STAR or click on the Edit Variable Information icon in the Data Viewer tool bar.

Data Manipulation 7 | Page

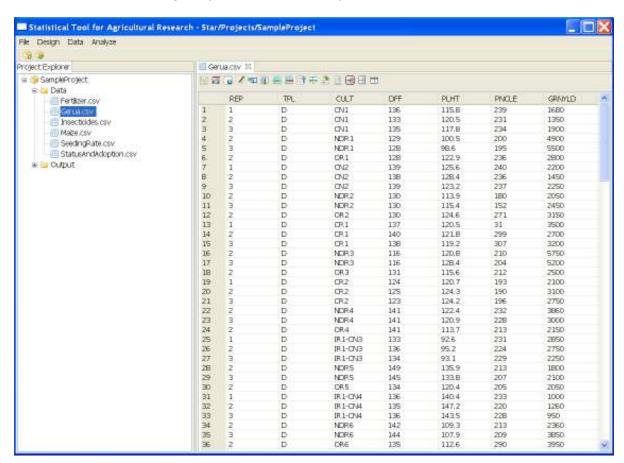
- The user can modify the variable type of the variables in the data set.
- Click OK.

3.7. Sorting

Re-arranging or sorting the rows or cases of the data file is often useful and sometimes necessary for certain types of analysis. The **Sort Cases Menu** can be used to sort rows based on the value of one or more sorting variables.

The steps to re-arrange the data set based on the value of one or more sorting variable are listed below:

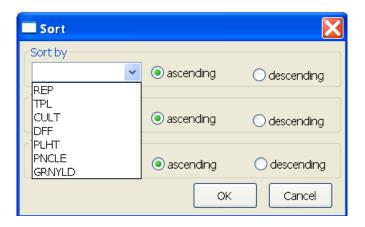
 On the Project Explorer, locate the dataset Gerua.csv from the Data folder of the project named SampleProject. Double-click the file to open it and view it in the Data Viewer. The data file is arranged by TPL, CULT then by REP.



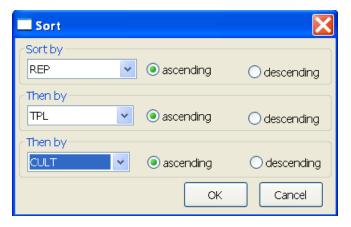
- Click Data | Sort from the main window of STAR or click on the Sort icon in the Data Viewer tool bar. The Sort dialog box will appear.
- The user can specify up to three variables as the basis for sorting and the order of sorting.

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Click the drop-down list box to identify the sorting variables. The rows can be rearranged or sorted in ascending or descending order, with ascending order as the default option. If multiple variables are selected, rows are sorted for each variable within categories of the preceding variables. For character variables, uppercase letters precede their lowercase counterparts. Further, the rows can be re-arranged or sorted in ascending or descending order, with ascending order as the default option.

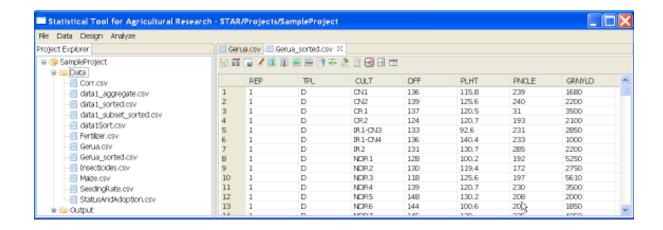


Suppose we want to re-arranging the content of the data *Gerua.csv*, by *REP*, *TPL* and *CULT*. The completed dialog box should appear as illustrated below:



Click the Ok button to sort the active data. The Sort dialog box is closed and STAR activates the new dataset in the Data Viewer tab.

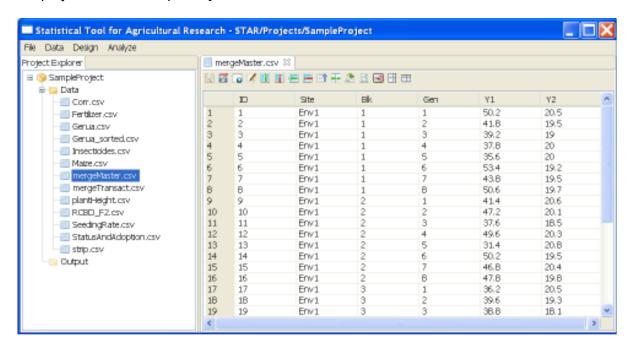
Data Manipulation 9 | Page



3.8. Merge Datasets

The **Merge submenu** can be used to combine the active data file (known as the master data) with another data file (referred as the transaction data) that contains the same cases or rows but different variables.

• On the Project Explorer, locate the *mergeMaster.csv* file from the *Data* folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.



- Choose Data | Merge Datasets or click on the Merge Datasets icon in the Data Viewer tool bar.
- Specify the required fields and appropriate options.

Transaction File Name

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Specify this file by selecting from the project *Data* folder using the drop-down combo box or by locating it using the **Browse** button. For the example, select *data2_merge.csv* file from the drop-down combo box.

Text File Delimeter

If the transaction file selected is a text file, the delimeter should be specified.

Observations to Include

The options available are to include common observations, all observations in the active data, all observations in the transaction data and all observations.

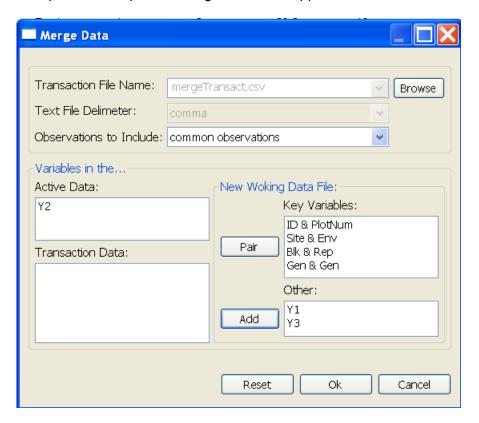
Key Variables

These are pair(s) of variables that are present in the active and transaction files and will be the basis for the merging. Variables can be paired by clicking one variable in the active data and clicking one variable in the transaction data then clicking the **Pair** button.

Other

These are the other variables the user wants to include in the merge data set.

For the example, the completed dialog box should appear as illustrated below:



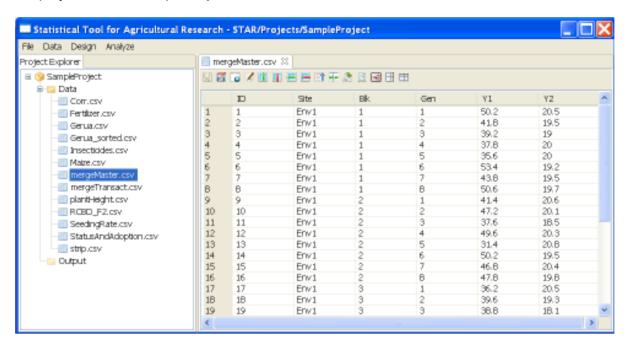
• Click OK. The merged data is saved in the *Data* folder and displayed in the Data Viewer.

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3.9. Append Datasets

The **Append submenu** can be used combine two data files with the same variables but different cases.

• On the Project Explorer, locate the *mergeMaster.csv* file from the *Data* folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.



- Choose Data | Append Datasets or click on the Append Datasets icon in the Data Viewer tool bar.
- Specify the required fields and appropriate options.

Transaction File Name

Specify this file by selecting from the project *Data* folder using the drop-down combo box or by locating it using the **Browse** button. For the example, select *data2_merge.csv* file from the drop-down combo box.

Text File Delimeter

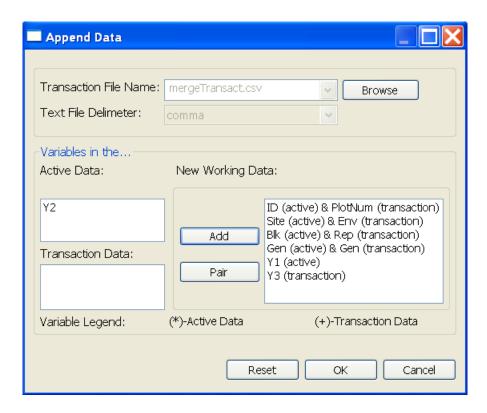
If the transaction file selected is a text file, the delimeter should be specified.

New Working Data

This is the list of variables or paired variables that will be included in the new data set. Variables can be paired by clicking one variable in the active data and clicking one variable in the transaction data then clicking the **Pair** button. Other variables can be added to this list by clicking the variable then clicking the **Add** button.

For the example, the completed dialog box should appear as illustrated below:

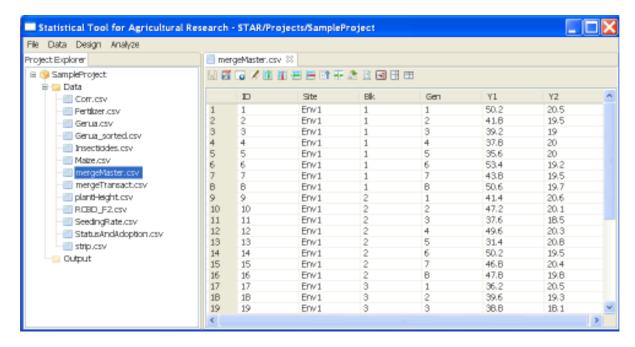
Data Manipulation 12 | Page



Click OK. The new data is saved in the Data folder and displayed in the Data Viewer.

3.10. Creating Data Subset

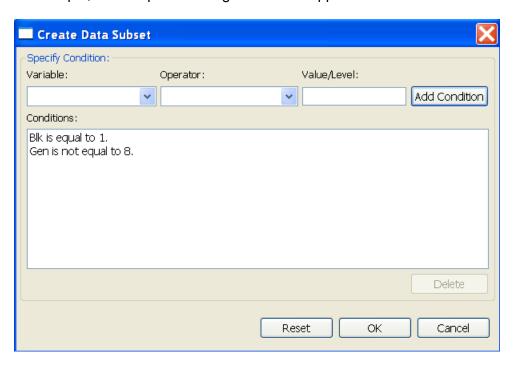
• On the Project Explorer, locate the *mergeMaster.csv* file from the Data folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.



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- Choose Data | Create Data Subset or click on the Create Data Subset icon in the Data Viewer tool bar.
- Suppose the user wants to select only the observations from Env2. This can be done by selecting variable Site, operator equals (==), typing Env2 under Value/Level and clicking the Add Condition button. Do the same for the condition "Gen not equal to 8".

For the example, the completed dialog box should appear as illustrated below:



Click OK. The new data is saved in the Data folder and displayed in the Data Viewer.

3.11. Reshaping

Data files are not always organized in the ideal form for specific needs of the users. The **Reshape subenu** can be used to restructure or reshape your active data set to follow the structure of the data set appropriate for the STAR procedure you want to use. This menu can also be used to combine data files, change the unit of analysis by grouping cases together or select a subset of observations to be used for the analysis.

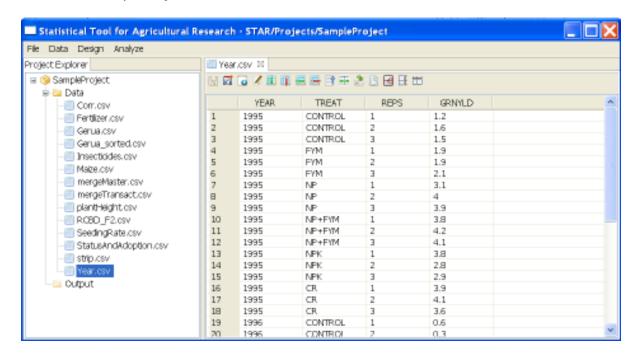
3.11.1. Reshaping Data from Long (Serial) to Wide (Parallel) Format

Reshaping data from long (serial) to wide (parallel) involves re-arranging a data file, such that, repeated measurements are in separate columns.

The steps for reshaping the data from long to wide format are listed below:

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• On the Project Explorer, locate the *Year.csv* file from the Data folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.



- Click Data | Reshape | Long (Serial) to Wide (Parallel) Format from the main window of STAR or click on the Rshape to Wide icon in the Data Viewer tool bar. The Manipulate Data: from Long (Serial) to Wide (Parallel) dialog box will automatically appear as shown below.
- Specify the required fields and appropriate options.

Variable(s) to be Reshape

This pertains to the list of variable(s) which is to be divided into two or more columns and saved to different variables in terms of the levels of the index factor.

Variables(s) to be Retained

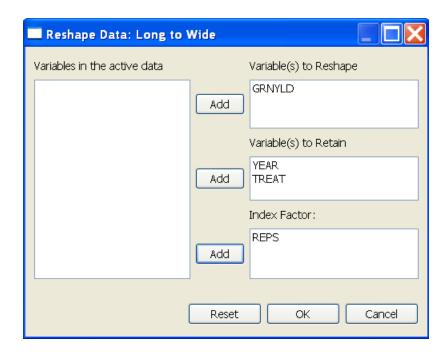
This pertains to variable(s) to be retained to describe the individual cases or observations.

Index Factor

This pertains to variable(s) which will determine the groupings of the values of the variable(s) to be transposed

For the example, the completed **Reshape Data: Long to Wide** dialog box should now look like this:

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 Click the Ok button to save the restructured data set. The new data is saved in the Data folder and displayed in the Data Viewer tab.

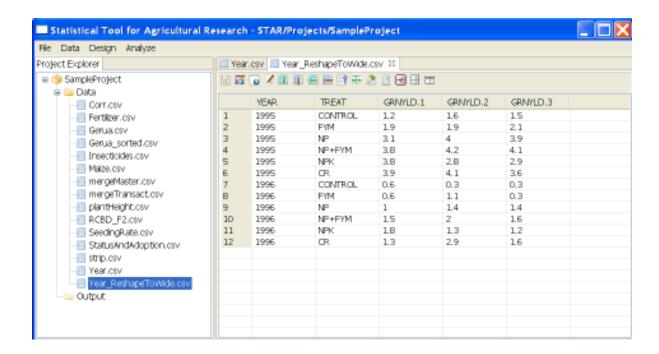
3.11.2. Reshape Data from Wide to Long Format

Manipulating data from wide (parallel) to long (serial) refers to re-arranging a multivariate into a univariate data. For instance, different columns representing measurements taken for a response variable over a period of time can be combined to form a single column, with the time variable used as an additional identifier variable.

The steps for reshaping the data from wide to long format are listed below:

 On the Project Explorer, locate the Year_ReshapeToWide.csv file (the generated file from section 3.11.1) from the Data folder of the project named SampleProject. Doubleclick the file to view it in the Data Viewer.

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- Click Data | Reshape | From Wide (Parallel) to Long (Serial) Format from the main window of STAR or click on the Reshape to Long icon in the Data Viewer tool bar. The Reshape Data: Wide to Long dialog box will appear.
- Specify the required fields and appropriate options,

Variable(s) to be Reshape

This pertains to variables which are to be combined into a single column.

Variable(s) to be Retained

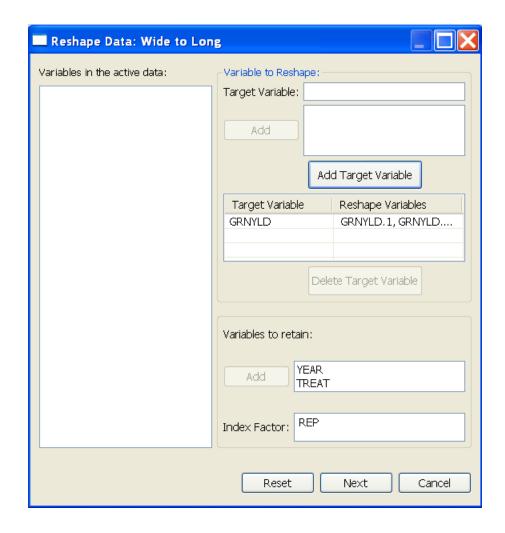
This pertains to variable(s) that describe the individual cases.

Index Factor

This is the new variable which will be create that pertains to the column where the particular observation came from.

For the example, the completed dialog box should appear as illustrated below:

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Click Ok to save the restructured data set. It can now be viewed in the Data View tab.

3.12. Aggregate

The **Aggregate submenu** aggregates group of cases into single cases and creates a new, aggregated file or creates new variables in the working data file that contain the aggregated data. Rows are aggregated based on the value of one or more grouping variables.

The steps for reshaping the data from long to wide format are listed below:

- On the Project Explorer, locate the *Year.csv* file from the Data folder of the project named *SampleProject*. Double-click the file to view it in the Data Viewer.
- Choose Data | Aggregate or click on the Aggregate Data icon in the Data Viewer tool bar.

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Specify the required fields and appropriate options.

Variable(s) to Aggregrate

This is the list of variables whose values will be summarized using the selected functions.

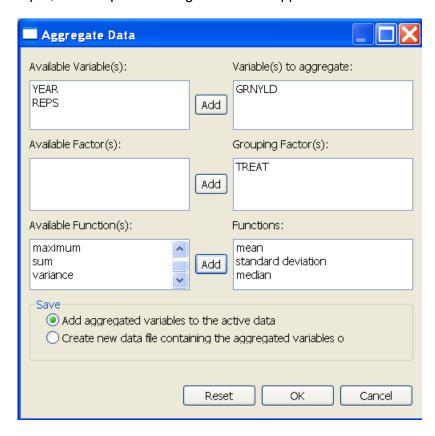
Grouping Factor(s)

Specify this grouping factor, if the user wants to summarize the values per level of a grouping factor.

Functions

Available functions are minimum, maximum, mean, median, sum, variance and standard deviation.

For the example, the completed dialog box should appear as illustrated below:



 Click OK. The active data is saved with the additional columns and displayed in the Data Viewer.

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4. Design Menu

The **Design menu** allows user to generate randomized field plans for single- and multifactor designs. This menu will generate a field book saved in *csv* format.

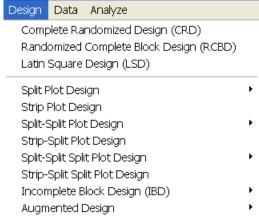


Figure 6 – 1. The Design Menu.

4.1. Completely Randomized Design (CRD)

To perform randomization, click **Design | Completely Randomized Design (CRD)**. The **Randomization** dialog box will appear.



The **Field Book Filename** text box lets the user specify the name of the data file that will be created. The default filename is *fieldbookCRD*.

Specify the required field. For the *example*, the completed dialog box should appear as illustrated below:



Sample Problem:

An experiment to compare the effects of six different rates of seeding (25, 50, 75, 100, 125,

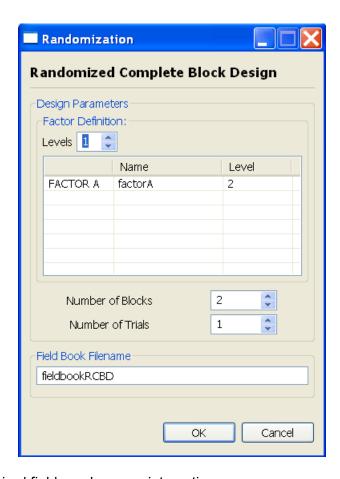
4.2. Single-Factor Randomized Complete Block Design (RCBD)

Suppose we want to generate a randomization for an experiment whose aim is to compare the effects of six different rates of seeding on grain yield which will be conducted in an RCBD with four replicates.

The steps to generate randomization for a Single-Factor in Randomized Complete Block Design are listed below:

Click Design | Randomized Complete Block Design (RCBD) from the main window.
 The Randomization dialog box will appear.

Design Menu 21 | Page



Specify the required fields and appropriate options.

Levels Spinbox

Define the number of factors. The number of factors specified will define the number of rows of the table inside the **Factor Definition** frame.

Factor Definition Frame

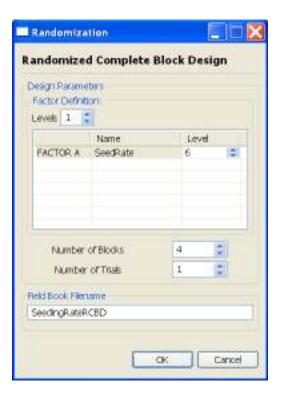
This table is required. It lets the user define the name of the factor(s) and its levels.

Field Book Filename Textbox

This field is required and it specifies the name where the data file will be created. The default filename is *fieldbookRCBD*.

For this example, the completed dialog box should appear as illustrated below:

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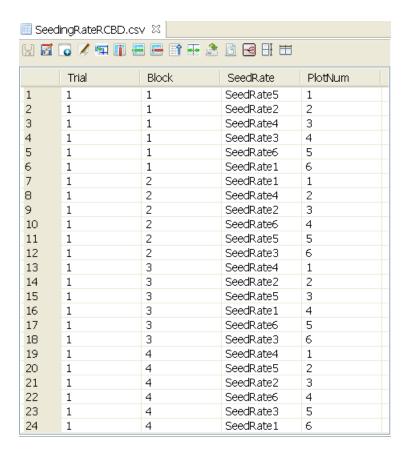


 Click the **OK** button to generate the randomization. The **Randomization** dialog box will be minimized and STAR open the data file created.



Sample data file that was created is shown below:

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4.2.1. Two-Factor Randomized Complete Block Design (RCBD)

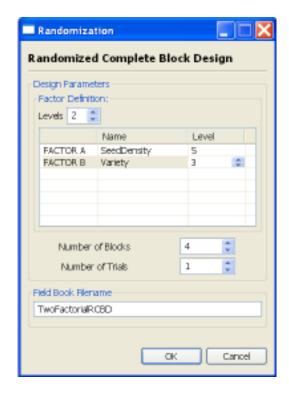
Suppose we want to generate a randomization for an experiment whose aim is to compare the effects of five seeding densities on the on grain yield of three rice varieties which will be conducted in an RCBD with four replicates

The steps to generate randomization for a Two-Factor in Randomized Complete Block Design are listed below:

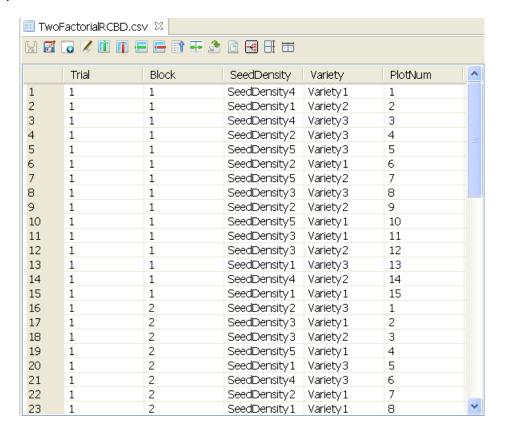
- Click Design | Randomized Complete Block Design (RCBD) from the main window.
 The Randomization dialog box will appear.
- Specify the required fields and appropriate options

For this example, the completed dialog box should appear as illustrated below:

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- Click the **OK** button to generate the randomization. The **Randomization** dialog box will be minimized and STAR open the data file created.
- Sample data file created is shown below:



Design Menu 25 | Page

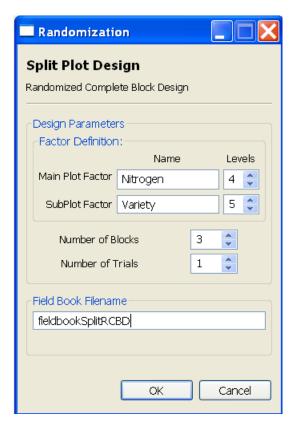
4.3. Split-Plot Design in RCBD

Suppose we want to generate a randomization for an experiment whose aim is to compare the effects of four nitrogen rates on the on grain yield of five rice varieties which will be conducted in RCBD using Split Plot with four replicates

The steps to generate randomization for Split Plot Design in Randomized Complete Block Design are listed below:

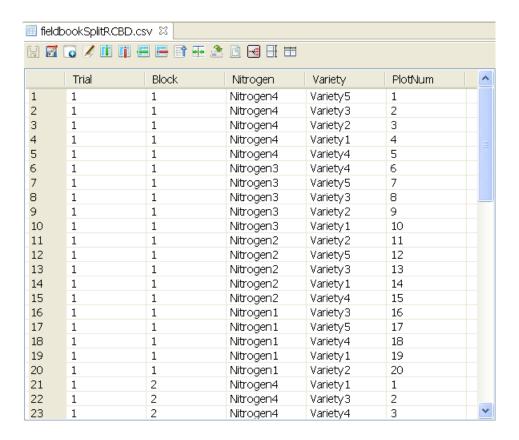
- Click Design | Split Plot Design | Randomized Complete Block Design (RCBD) from the main window. The Randomization dialog box will appear.
- Specify the required field and appropriate options.

For this example, the completed dialog box should appear as illustrated below:



- Click the **OK** button to generate the randomization. The **Randomization** dialog box will be minimized and STAR actives the Data Viewer tab.
- Sample data file created is shown below:

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4.4. Strip-Plot Design

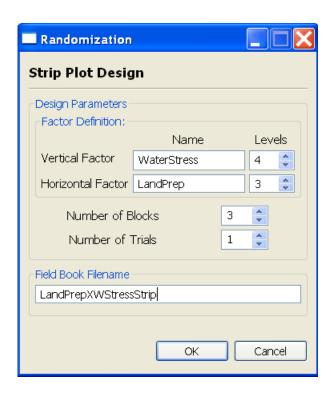
Suppose we want to generate a randomization for an experiment to be conducted using Strip Plot, with land preparation (with 3 levels) as the horizontal factor and water stress (with four levels) as the vertical factor. Three replicates will be used.

The steps to generate randomization for a Strip Plot Design are listed below:

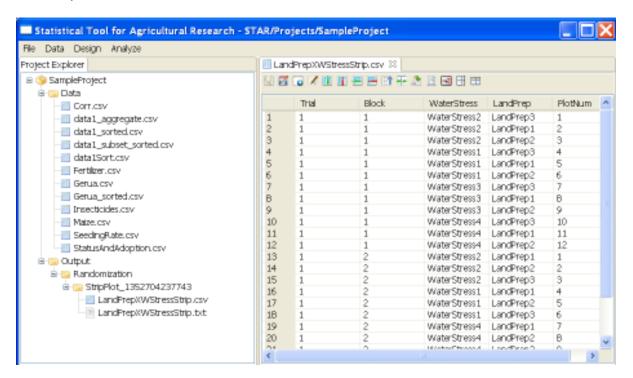
- Choose Design | Strip Plot Design from the main window. The Randomization dialog box will appear.
- Specify the required fields and appropriate options.

For this example, the completed dialog box should appear as illustrated below:

Design Menu 27 | Page



- Click the OK button to generate the randomization. The Randomization dialog box will be minimized and STAR actives the Data Viewer tab.
- Sample data file created is shown below:



Design Menu 28 | Page

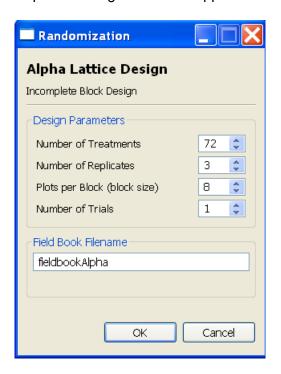
4.5. Alpha Lattice Design

Suppose we want to generate a randomization for an experiment that will be conducted using Alpha Lattice design with 72 entries grouped into 8 with three replicates.

The steps to generate randomization for Alpha Lattice Design are listed below:

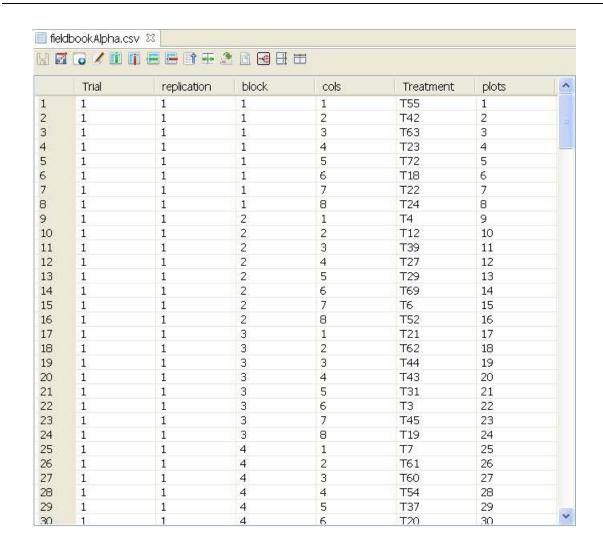
 Click Design | Incomplete Block Design | Alpha Lattice Design from the main window. The Randomization dialog box will appear.

For this example, the completed dialog box should appear as illustrated below:



- Click the **OK** button to generate the randomization. The **Randomization** dialog box will be minimized and STAR actives the Data Viewer tab.
- Sample data file created is shown below:

Design Menu 29 | Page



4.6. Augmented Design in RCB

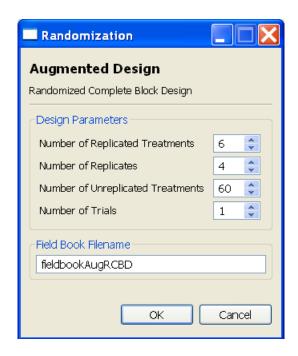
Suppose we want to generate a randomization for an experiment which will be conducted using Augmented design in RCB involving four replicates. The experiment will use six replicated treatment and 60 unreplicated treatment (test entries).

The steps to generate randomization for a Augmented Design in Randomized Complete Block are listed below:

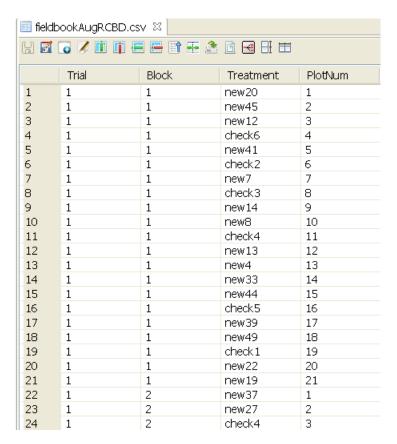
Click Design | Augmented Design | Randomized Complete Block Design (RCBD) from the main window. The Randomization dialog box will appear.

For this example, the completed dialog box should appear as illustrated below:

Design Menu 30 | Page



- Click the **OK** button to generate the randomization. The **Randomization** dialog box will be minimized and STAR actives the Data Viewer tab.
- Sample data file created is shown below:



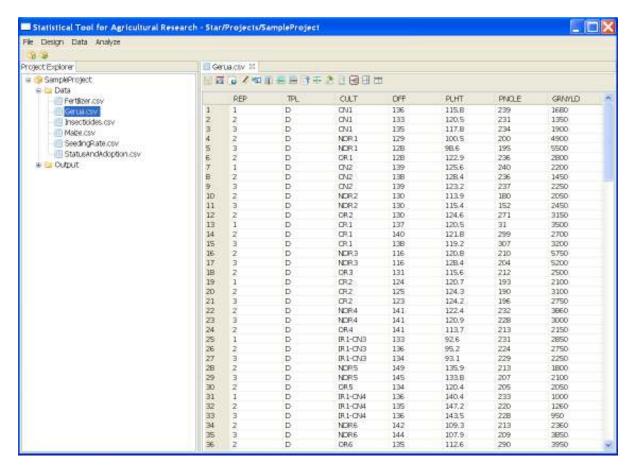
Design Menu 31 | Page

5. Descriptive Statistics

The **Descriptive submenu** of STAR performs univariate summary statistics for several variables and/or per levels of grouping variables.

The steps to perform Descriptive are listed below:

• On the Project Explorer, locate the dataset *Gerua.csv* from the *Data* folder of the project named *SampleProject*. Double-click the file to open it and view it in the Data Viewer.



- To generate descriptive statistics, from the main window click **Analyze | Descriptive Statistics | Descriptive**. The **Descriptive Statistics** dialog box will appear.
- Specify the required field and appropriate options for the analysis.

Variable Description Tab

Variable(s)

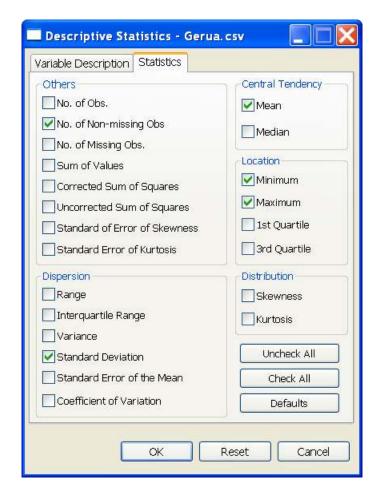
This field is required and at least one entry is need for the analysis to proceed.

By Variable(s)

This field is optional. If a **By Variable(s)** is specified, it will generate descriptive statistic per category of the **By Variable**.

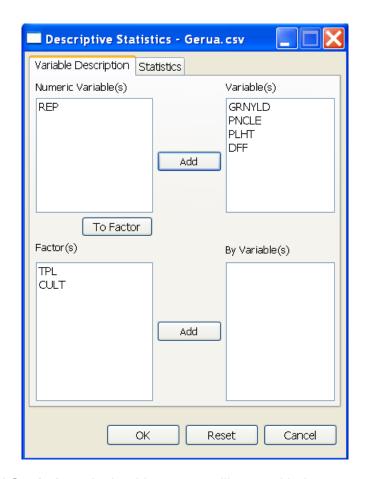
Statistics Tab

This tab lets the user identify the numerical descriptive measures to be displayed. By default, the procedure will display five statistics, namely: number of non-missing observations (NNMissObs), minimum (min), maximum (max), mean, and standard deviation (stdev). To request for all statistics, click the **Check All** button of the **Statistics** tab. You may click on the **Uncheck All** button to remove previously selected items and start with a new selection.



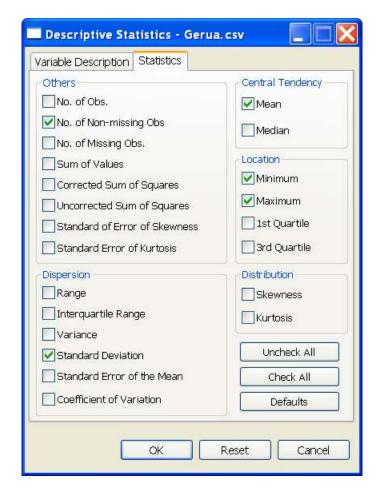
Suppose we want to compute the descriptive statistics for variables *DFF*, *PLHT*, *PNCLE* and *GRNYLD*. The completed **Variable Description** tab should appear as illustrated below:

Descriptive Statistics 33 | Page



The completed Statistics tab should appear as illustrated below:

Descriptive Statistics 34 | Page

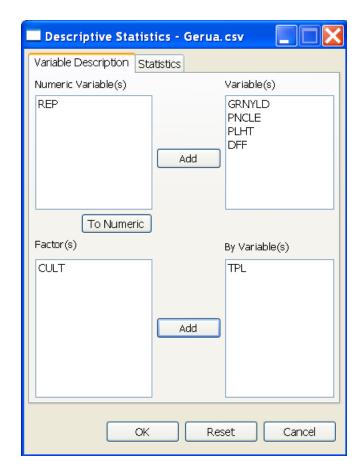


- Click the Ok button to generate the summary statistics. The Descriptive Statistics dialog box will be minimize and STAR activates the Output Viewer tab.
- Sample output of the descriptive statistics is shown below:

DESCRIPTIV	E STATISTICS				
Variable	N_NonMissObs	Min	 Max	Mean	StdDev
GRNYLD PNCLE PLHT DFF	156 156 156 156	950.00 29.00 92.60 14.00	6500.00 336.00 163.50 151.00	3759.20 230.12 128.48 125.26	1323.36 46.19 15.29 14.65

 Suppose we want to compute the descriptive statistics for variables DFF, PLHT, PNCLE and GRNYLD per level of TPL. The completed Variable Description tab should appear as illustrated below:

Descriptive Statistics 35 | Page



Click the **Ok** button to generate the summary statistics. Sample output of the descriptive statistics is shown below:

DECCD	TPTTVF.	CHVH.	COTTCC

Variable	TPL	N_NonMissObs	Min	Max	Mean	StdDev
GRNYLD	D	78	950.00	5750.00	2919.68	1226.98
GRNYLD	N	78	2900.00	6500.00	4598.72	769.37
PNCLE	D	78	31.00	307.00	212.19	42.49
PNCLE	N	78	29.00	336.00	248.05	42.87
PLHT	D	78	92.60	149.30	120.05	11.38
PLHT	N	78	115.20	163.50	136.90	14.05
DFF	D	78	14.00	151.00	133.85	15.64
DFF	N	78	106.00	128.00	116.67	6.14

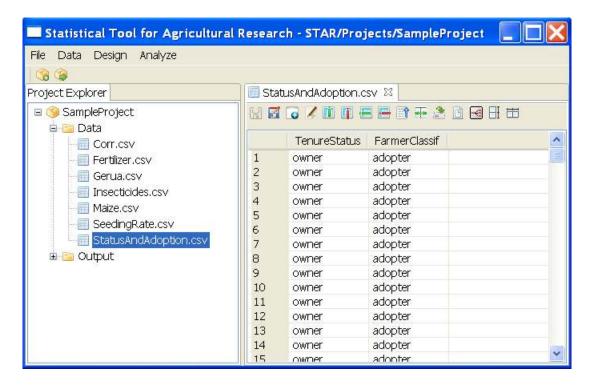
Descriptive Statistics 36 | Page

6. Cross Tabulation

The cross tabulation forms two-way tables and provides a variety of tests and measures of association for a two-way tables.

The steps to perform Cross Tabulation are listed below:

 Locate the dataset StatusAndAdoption.csv from the Data folder of the project named SampleProject in the Project Explorer. Double-click the file to open it and view it in the Data Viewer.



- To generate cross table, from the main window click Analyze | Descriptive Statistics |
 Cross Tabulation. The Cross Tabulation dialog box will appear.
- Specify the required field and the appropriate options for the analysis.

Variable Description Tab

For the analysis to proceed, the **Row Variable(s)** list box and the **Column Variable(s)** list box from the **Variable Description** tab should both have at least one entry and at least one checkbox should be tick from the **Option** tab.

Options Tab

For the analysis to proceed, at least one checkbox should be tick in this tab.

Cell Display

The Crosstabs procedure displays the observed and expected frequency. Each cell of the table may contain counts and percentages.

Counts

Display the number of observed frequency and number of expected frequency. By default, the observed frequency is displayed.

Percentages

This field is optional. It may display the row, column or the total percentages.

Residuals

Statistics

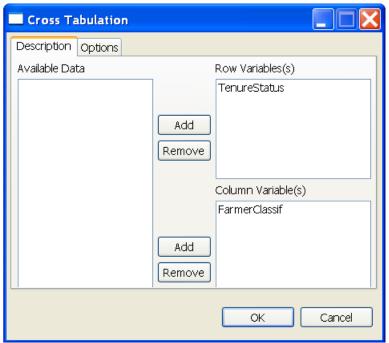
Chi-square

Select Chi-square to calculate the Chi-square Test of Independence and the likelihood-ratio chi-square.

Nominal

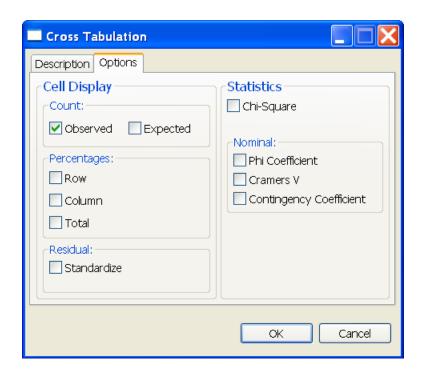
For nominal data, three measure of association based on chi-square (Phi Coefficient, Cramer's V and Contingency Coefficient) may be requested.

For the example, the completed **Variable Description** tab should appear as illustrated below:



The completed **Options** tab should appear as illustrated below:

Cross Tabulation 38 | Page



Click the Ok button. Sample output of the analysis using the default option is shown below:

CROSS TABULATION

Table of Observed Frequency: TenureStatus by FarmerClassif

	FarmerClassif		
	adopter		Total
fixed-rent	4	3	7
owner	102	26	128
share-rent	42	10	52
Total	148	39	187

Sample output of the analysis if all checkbox is tick is shown below:

CROSS TABULATION

Table of TenureStatus by FarmerClassif

		FarmerClassif	+	.
TenureStatus		 adopter +	 nonadopter 	 Total
fixed-rent	Obs Freq Expected Freq	4 5.5401 +		7 7 7 7 7 7 7 7 7 7

Cross Tabulation 39 | Page

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owner	Obs Freq Expected F	 Freq	102	- 1	128
share-rent	Obs Freq Expected F	req	42 41.1551	10.8449	52
Total	Obs Freq Expected F	req	148.0000	39.0000	187.0000

Statistics for Table TenureStatus by FarmerClassif

Statistics	DF	Value	Prob
Pearson Chi-Square	2	2.1589	0.3398
Likelihood Ratio Chi-Square	2	1.8236	0.4018
Phi Coefficient		0.1074	
Contingency Coefficient		0.1068	
Cramer's V		0.1074	

^{*} Cells with Expected Frequency < 5: 2 of 6 (33.33%)

Cross Tabulation 40 | Page

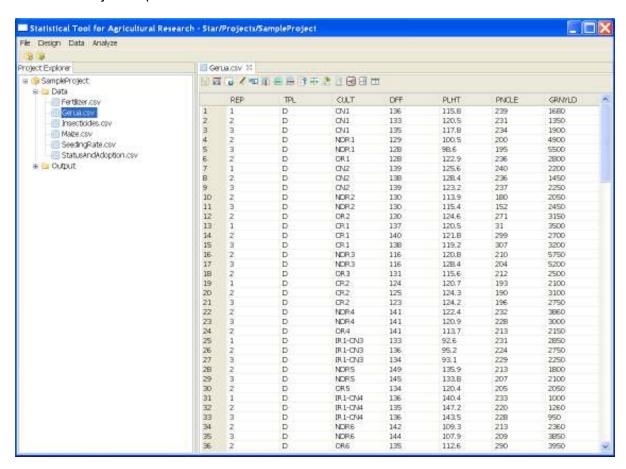
7. T-test

7.1. One Sample t-tests

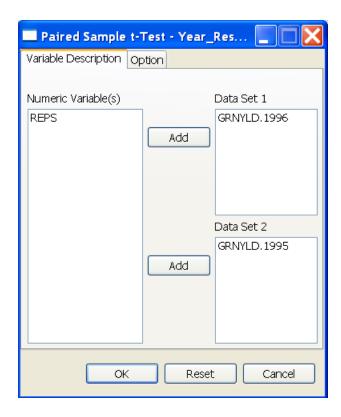
The One Sample *t*-test procedure determine whether the mean of one variable differ from the hypothesized value. It assumes that the samples are randomly taken from the population of interest and that the observations are obtained from a normal distribution.

The steps to perform Heterogeneity test are listed below:

• Locate the dataset *Gerua.csv* from the *Data* folder of the project named *SampleProject* in the Project Explorer.



- Click Analyze | t-Test | One Sample. The One Sample dialog box will appear.
- Opening the data for the first time, Rep field is regarded by R as numerical variables, they need to be changed as factors. Choose the variable and click on the To Factor button.
- Specify the required field and appropriate options for the analysis.



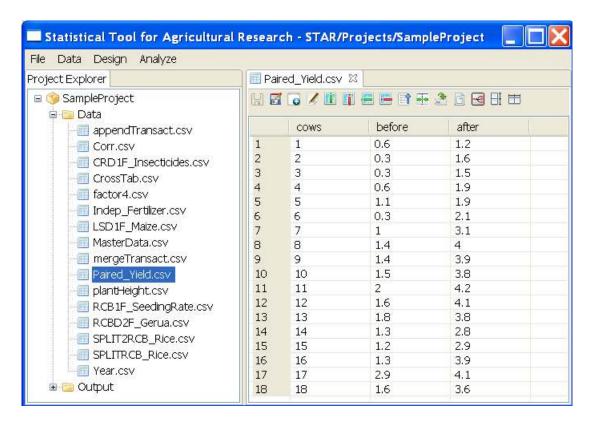
7.2. Paired Sample t-tests

The paired sample *t*-test is used to compares the means of paired or related samples. It assumes that the samples are randomly taken from the population of interest and the population of differences is normally distributed.

The steps to perform paired t-test are listed below:

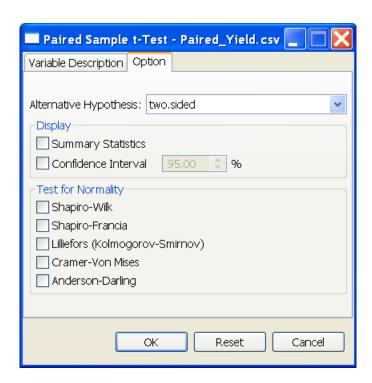
 Locate the dataset Paired_.csv from the Data folder of the project named SampleProject in the Project Explorer.

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- Click Analyze | t-Tests | Paired Samples. The Paired Samples dialog box will appear.
- Specify the required field and appropriate options for the analysis.

Options Tab



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Alternative Hypothesis

User can select a one-tailed or two-tailed test. The default alternative hypothesis used is that for the two-sided test.

Level of Significance

This will be used to determine if the two groups have equal variance. Its default value is 0.05. The user can change this value by specifying a numeric value from zero to 1.

Display Summary Statistics

If this option is selected, a summary table with number of observations, minimum, maximum, mean and standard deviation will be displayed.

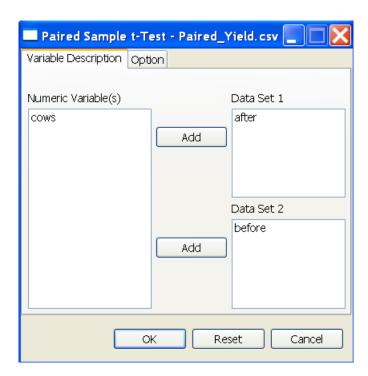
Display Confidence Interval

If this option is selected, a confidence interval of the mean will be displayed. If this is selected, by default, a 95% confidence interval is displayed. Valid value is between 90% to 99%.

Test Procedure

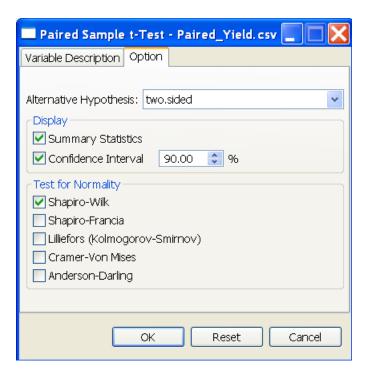
If at least one option is selected, if display the test for normality test for each level of the **Grouping Variable**. There are five available test for normality procedure.

For the example, the completed **Variable Description** tab should appear as illustrated below:



The completed **Option** tab should appear as illustrated below:

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- Click the **OK** button to perform the analysis.
- Sample output of the analysis is shown below.

Test for Normality								
Difference	Meth	od	St	at	Value	p Va	lue	
after - before	Shap	iro- <u>Wilk</u>	M	C	.9420	0.3		
Descriptive Sta								
Difference								_
after - before		1.5	3	1.79		2.04		0.1468
* At 90% Confid	ence							
PAIRED SAMPLE t					_			
Difference					-			
after - before	17 	12.18		0.000C	-) -			

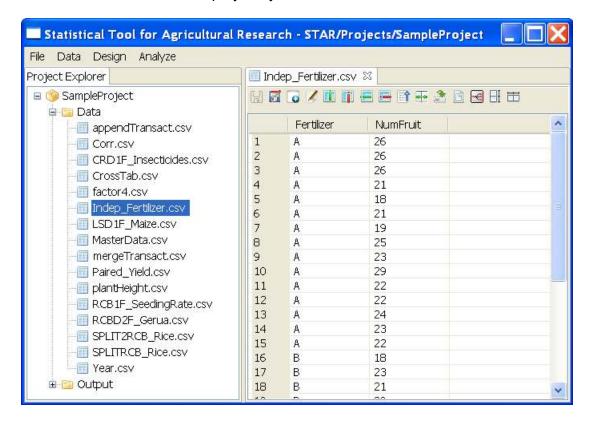
7.3. Independent Sample *t*-tests

t-Test 45 | Page

The independent sample *t*-test is used to determine whether the means between two groups or populations are equal. It assumes that (i) the samples are drawn independently from each population; (ii) the samples from each population are randomly taken; and (iii) the observations from each population are obtained from a normal distribution.

The steps to perform *t*-Test for two independent samples are listed below:

On the Project Explorer, locate the dataset Indep_Ferlitizer.csv from in the Data folder of the project named SampleProject. Double-click the file to open and view it in the Data Viewer. The file contains the number of fruits (NumFruit) from trees sprayed by two fertilizers (Fertilizer – A and B). Suppose we want to test the hypothesis that the mean numbers of fruits from trees sprayed by the two fertilizers are the same.



To perform t-test on two independent samples, from the main window of STAR click Analyze | t-test | Independent Sample. The Two Independent Sample dialog box will appear.

Variable Description Tab

Test Variable(s)

This field is required for the analysis to proceed. This field need at least one entry and should come from the **Numeric Variable(s)** list box.

Grouping Variable(s)

This field is required and will only accept one entry at a time. The entry must came from the **Factor(s)** list box. The entry should have only two levels.

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Options Tab

Alternative Hypothesis

User can select a one-tailed or two-tailed test. The default alternative hypothesis used is that for the two-sided test.

Level of Significance

This will be used to determine if the two groups have equal variance. Its default value is 0.05. The user can change this value by specifying a numeric value from zero to 1.

Display Summary Statistics

If this option is selected, a summary table with number of observations, minimum, maximum, mean and standard deviation will be displayed.

Display Confidence Interval

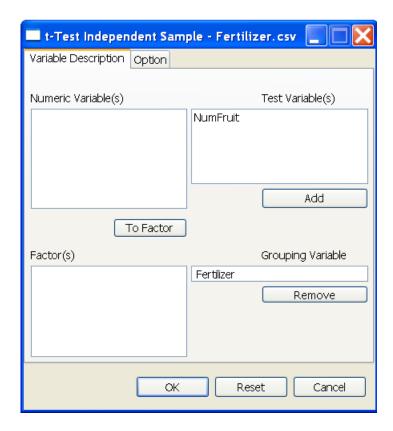
If this option is selected, a confidence interval of the mean will be displayed. If this is selected, by default, a 95% confidence interval is displayed. Valid value is between 90% to 99%.

Test Procedure

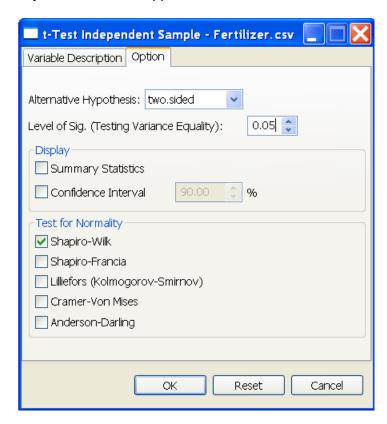
If at least one option is selected, if display the test for normality test for each level of the **Grouping Variable**. There are five available test for normality procedure.

For the example, the completed **Variable Description** tab should appear as illustrated below:

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The completed **Option** tab should appear as illustrated below:



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- Click the **OK** button to perform the analysis.
- Sample output of the analysis is shown below. The default output of the t-Test for two
 independent samples includes the test for equality of variances. Depending on the result
 of the equality of variances, Pooled t-Test or a t-test with satterwaite adjustment is
 computed.

прсп	FOR	NORMALITY
TEOT		NORMALLI

_			Method			_
Fertilizer Fertilizer	A B	NumFruit NumFruit	Shapiro-Wilk Shapiro-Wilk	W W	0.9726 0.9198	0.8950 0.1916
Equality of						
Variable M	ethod	Num DF	Den DF F Val	ue Pr	> F	
NumFruit F	olded F	14	14 1.	47 0.4		
TWO INDEPEN	DENT SA	MPLE t-TES'				
			DF t Value		t	
			28 1.91		0663	

^{*} At 0.05 level of significance.

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8. Analysis of Variance

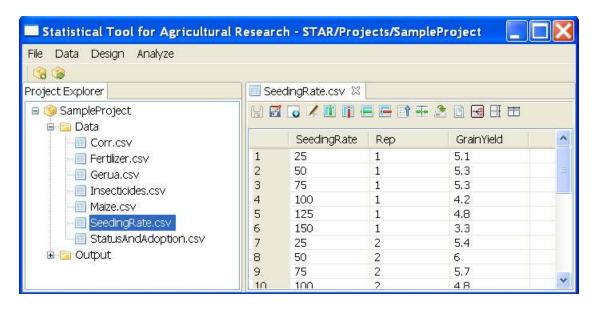
The **Analysis of Variance submenu** can be use to perform analysis of variance for data from experimental designs with single error (e.g., RCB) and with 2 or more errors (e.g., Split Plot).

Pairwise mean comparisons and contrast analyses are available to evaluate differences among specific treatment/treatment combination.

8.1. One Factor Randomized Complete Block

The steps to perform the Analysis of Variance in Randomized Complete Block Design are listed below:

• On the Project Explorer the dataset SeedingRate.csv located in the Data folder of the project named SampleProject. Double-click the file to open and view it in the Data Viewer. The example file contains data from an experiment conducted to investigate the effect of six different rates of seeding (SeedingRate, in kg seed/ha with levels 25, 50, 75, 100, 125 and 150) on the grain yield (GrainYield, in kg/ha) of rice a variety using Randomized Complete Block (RCB) field design with four blocks (Rep with levels 1, 2, 3 and 4).



- From the main window of STAR click Analyze | Analysis of Variance | Randomized Complete Block Design (RCBD). The Randomized Complete Block Design (RCBD) dialog box will appear.
- Opening the data for the first time, SeedingRate and Rep fields in the data file are regarded by R as numerical variables; they need to be changed as factors. Choose the variable and click on the To Factor button.
- Specify the required field and appropriate options for the analysis.

Model Specification Tab

Response Variable(s)

This field is required. For the analysis to proceed, at least one item should be entered in the list box. Only Items from the **Numeric Variable(s)** can be added in this dialog box.

Factor(s)

This field is required. At least one item can be added in this list box and should come from the **Factor(s)** list box.

Block

This field is required. Only one item can be added in this field and should come from the **Factor(s)** list box.

Options Tab

<u>Display Descriptive Statistics</u>

If this option is selected, a summary table with number of non-missing observations, minimum, maximum, mean and standard deviation of the response variable will be displayed.

Shapiro-Wilk Test

If this option is selected, test for normality of residuals using Shapiro-Wilk will be displayed.

Bartlett's Test

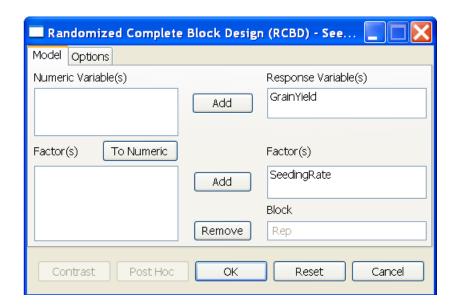
If this option is selected, test for homogeneity of variances using Bartlett's Test will be displayed.

Significance Level

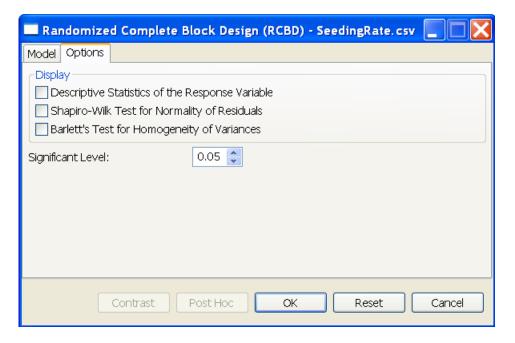
This will be used to determine if the treatment effects and interaction effects are significant in Analysis of Variance. This is also the level of significance to be used when pairwise mean comparison is performed. Its default value is 0.05. The user can change this value by specifying a numeric value from zero to 1.

For the example, the completed **Model** Specification tab should appear as illustrated below:

Analysis of Variance 51 | Page



The completed **Options** tab should appear as illustrated below:



 Click OK button to perform the analysis. The Randomized Complete Block Design (RCBD) dialog box will be minimized and STAR activates the Output Viewer tab and the Graph Viewer tab.

Output Viewer

This tab contains the text output which begins with the descriptive statistics and the output for the tests for homogeneity of variances and test for normality, whichever was specified in the **Options** tab. The next part is the Analysis of Variance table. If there is only one factor considered in the analysis, and if the factor is not significant a table of treatment means will be displayed. If there are 2 or more factors in the analysis and if

Analysis of Variance 52 | Page

the highest interaction is not significant, table of means for all levels of the treatment combinations will be displayed. Then, depending on the result of the analysis of variance, pairwise comparison of means will be automatically generated using one applicable pairwise comparison procedure. If at most 5 treatment levels are to be compared, the default pairwise comparison procedure used in STAR is the Least Significant Difference (LSD) test; otherwise, the Honestly Significant Difference (HSD) or Tukey's test is used.

Graph Viewer

The tab displays two diagnostics plots. The left plot, entitled *Residuals vs. Fitted* is for assessing the distribution of the residuals. An ideal plot should reveal that the points are distributed around 0. The residuals should not be affected by the size of the fitted values. Problematic data may result to a plot wherein the points increase as the fitted values increase, thus forming a "funnel-shaped" distribution. This will indicate a violation of the homogeneity of variances requirement for analysis of variance. The *Normal Q-Q* plot is used for assessing normality. The points should lie about the line.

Sample text output of the analysis displayed in the Output Viewer tab is shown below:

```
Analysis of Variance
Randomized Complete Block Design (RCBD)
CLASS INFORMATION
Class Level Information
______
        NO. OF LEVELS LEVELS
SeedingRate
               6 25, 50, ..., 150
                4 1, 2, 3, 4
.____
Number of Observations Read and Used: 24
Bartlett's Test for Homogeneity of Variances
_____
Method DF Test Stat p Value
bartlett 5 4.76 0.4462
 .____
TEST FOR NORMALITY
      Method Stat Value p Value
GrainYield resid Shapiro-Wilk W 0.9346 0.1235
_____
ANOVA TABLE
Response Variable: GrainYield
      DF Sum of Square Mean Square F Value Pr > F
_____
             1.7667 0.5889 4.87 0.0147
         3
```

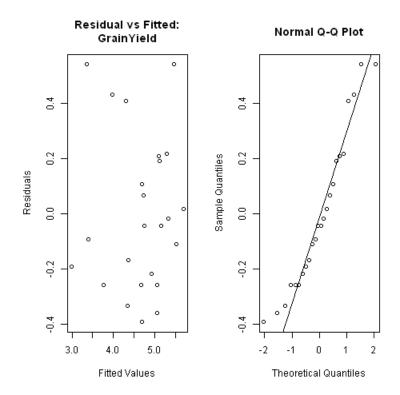
Analysis of Variance 53 | Page

SeedingRate Error Total	15	1.81 13.76	33 00	0.1209	16.84	0.0000
Summary State						
Coef Var G	rainYield	Mean				
7.48		4.65				
Standard Erro						
Effects	StdErr					
Rep SeedingRate	0.2007 0.2459					
Pairwise Mear	n Comparis	on of Se	edingRate	Э		
Tukeys's Hone	est Signif	icant Di	fference	(HSD) Tes	st	
Alpha Error Degrees Error Mean So Critical Valu Test Statisti	quare 1e	om	0.05 15 0.1209 4.5947 0.7988			
Summary of th						
SeedingRate	means N					
125 150 25 50 75	4.35 4	ab c ab ab a				

^{*} Means with the same letter are not significantly different.

Sample graphical output of the analysis displayed in the Graph Viewer tab is shown below:

Analysis of Variance 54 | Page



Additional Options

If there are still modifications to be made, the following option buttons may be used:

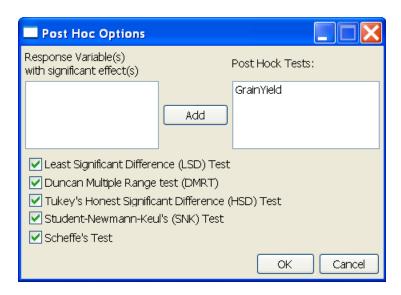
Post Hoc

Use this option if other pairwise comparison procedures, other than what was presented in the **Output Viewer** tab page, are desired. There are five pairwise mean comparison procedure available, namely: Least Significant Difference Test, Duncan Multiple Range Test, Tukey's Honest Significant Test, Student Newmann-Keul's Test and the Scheffe's Test. This button will be disabled if *F*-tests conducted using ANOVA yielded results that are not significant.

Suppose additional pairwise mean comparison is requested.

- o Click the **Post Hoc Option** button. The **Post Hoc Option** dialog box will appear.
- Specify the at least one response variable where pairwise mean comparison will be perform and choose at least one pairwise mean comparison procedure.
- The completed Post Hoc Option dialog box should appear as shown below:

Analysis of Variance 55 | Page



- Click the **OK** button to perform the pairwise mean comparison. The **Post Hoc Option** dialog box will be closed and STAR activates the **Output Viewer** tab.
- Sample result of the pairwise mean comparison requested is shown below:

```
Result of Pairwise Comparison
Response Variable: GrainYield
Pairwise Mean Comparison of SeedingRate
                        0.05
Alpha
Error Degrees of Freedom
                          15
                      0.1209
Error Mean Square
                 LSD HSD scheffe
Critical Value 2.1314 4.5947 2.9013
Test Statistic 0.5240 0.7988 0.9364
Summary:
SeedingRate means N std.err LSD HSD scheffe
            4.35 4 0.1708 c b
100
                                     b
125
           4.67 4 0.0946 bc ab ab
            3.38 4 0.2287 d c c
150
           5.12 4 0.1548 ab ab a
25
            5.08 4
                    0.3705 ab ab
50
            5.30 4 0.2160 a
                               a a
* Means with the same letter are not significantly different
Duncan's Multiple Range Test (DMRT)
                             0.05
Alpha
Error Degrees of Freedom
                               15
```

Analysis of Variance 56 | Page

Error Mean Square			0.1209		
Number of Means	2	3	4	5	
Tabular Value Test Statistics				3.31 0.58	

Summary of the Result:

SeedingRate	means	N	group
100	4.35	4	С
125	4.67	4	bc
150	3.38	4	d
25	5.12	4	ab
50	5.08	4	ab
75	5.30	4	a

^{*} Means with the same letter are not significantly different.

Student Newman Keuls (SNK) Test

Alpha		0.05
Error	Degrees of Freedom	15
Error	Mean Square	0.1209

Number of Means	2	3	4	5	6
Critical Value	3.0143	3.6734	4.0760	4.3670	4.5947
Test Statistics	0.5240	0.6386	0.7086	0.7592	0.7988

Summary of the Result:

SeedingRate	means	N	group
100	4.35	4	b
125	4.67	4	ab
150	3.38	4	С
25	5.12	4	a
50	5.08	4	a
75	5.30	4	a

^{*} Means with the same letter are not significantly different.

Contrast

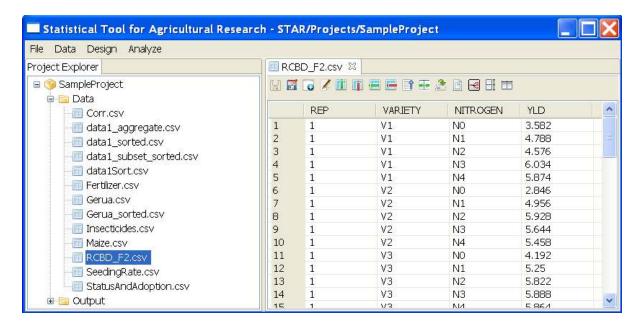
This button will be enabled if there is at least one factor that has more than two levels. Use of this button will be discussed in Partitioning Sum of Squares section of this module.

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8.2. Two Factor Randomized Complete Block

The steps to perform the Analysis of Variance in Randomized Complete Block Design are listed below:

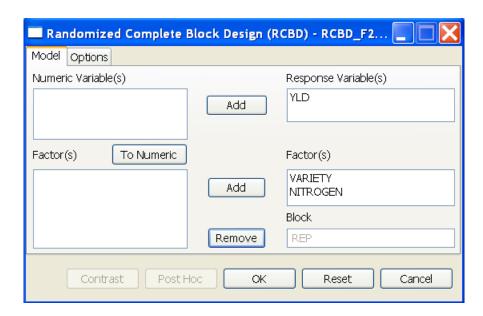
• On the Project Explorer the dataset RCBD_F2.csv located in the Data folder of the project named SampleProject. Double-click the file to open and view it in the Data Viewer. The example file contains data from an experiment conducted to compare the effects of three rice variety (VARIETY, with levels V1, V2 and V3) and five nitrogen levels (NITROGEN, with levels N0, N1, N2, N3 and N4) on grain yield (YLD, in t/ha) using Randomized Complete Block (RCB) field design with four blocks (REP with levels 1, 2, 3 and 4).



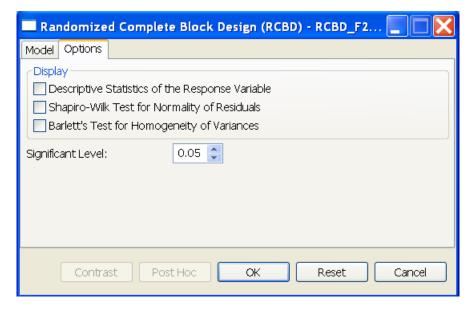
- From the main window of STAR click Analyze | Analysis of Variance | Randomized Complete Block Design (RCBD). The Randomized Complete Block Design (RCBD) dialog box will appear.
- Opening the data for the first time, REP field in the data file is regarded by R as numerical variable; it need to be changed as factors. Choose the variable and click on the To Factor button.
- Specify the required field and appropriate options for the analysis.

For the example, the completed **Model** Specification tab should appear as illustrated below:

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The completed **Options** tab should appear as illustrated below:



- Click OK button to perform the analysis. The Randomized Complete Block Design (RCBD) dialog box will be minimized and STAR activates the Output Viewer tab and the Graph Viewer tab.
- Sample text output of the analysis displayed in the Output Viewer tab is shown below:

```
Analysis of Variance
Randomized Complete Block Design (RCBD)

CLASS INFORMATION
Class Level Information
FACTOR NO. OF LEVELS LEVELS
```

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REP 4 1, 2, 3, 4 VARIETY 3 V1, V2, V3 NITROGEN 5 NO N1 N2 N

NITROGEN 5 NO, N1, N2, N3, N4

Number of Observations Read and Used: 60

ANOVA TABLE

Response Variable: YLD

Source	DF	Sum of Square	Mean Square	F Value	Pr > F
REP VARIETY NITROGEN VARIETY:NITROGEN Error Total	3 2 4 8 42 59	2.5018 1.1527 42.0064 2.3943 6.1157 54.1709	0.8339 0.5763 10.5016 0.2993 0.1456	5.73 3.96 72.12 2.06	0.0022 0.0266 0.0000 0.0627

Summary Statistics

Coef Var YLD Mean -----7.71 4.95

Standard Errors

Effects StdErr

REP 0.1393
VARIETY 0.1207
NITROGEN 0.1558
VARIETY:NITROGEN 0.2698

Table of Means

NITROGEN NO N1 N2 N3 N4
VARIETY
V1 3.0565 4.7235 4.4595 5.7170 5.8230
V2 3.5480 4.8060 5.4360 5.7345 5.6805
V3 3.7780 4.7535 5.3200 5.5525 5.8865

Pairwise Mean Comparison of VARIETY

Least Significant Difference (LSD) Test

Alpha 0.05
Error Degrees of Freedom 42
Error Mean Square 0.1456
Critical Value 2.0181
Test Statistics 0.2435

Summary of the Result:

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VARIETY	means	N	group	
				•
V1	4.76	20	b	
V2	5.04	20	a	
V3	5.06	20	a	

^{*} Means with the same letter are not significantly different.

Pairwise Mean Comparison of NITROGEN

Least Significant Difference (LSD) Test

Alpha	0.05
Error Degrees of Freedom	42
Error Mean Square	0.1456
Critical Value	2.0181
Test Statistics	0.3144

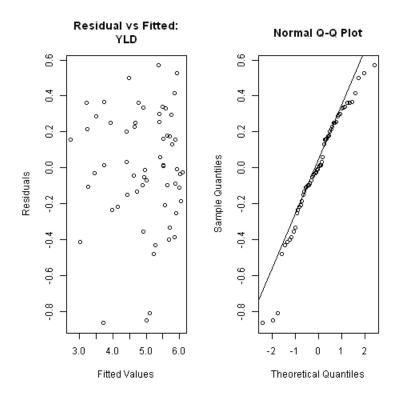
Summary of the Result:

				-
NITROGEN	means	N	group	
				-
NO	3.46	12	С	
N1	4.76	12	b	
N2	5.07	12	b	
N3	5.67	12	a	
N4	5.80	12	a	

^{*} Means with the same letter are not significantly different.

Sample graphical output of the analysis displayed in the Graph Viewer tab is shown below:

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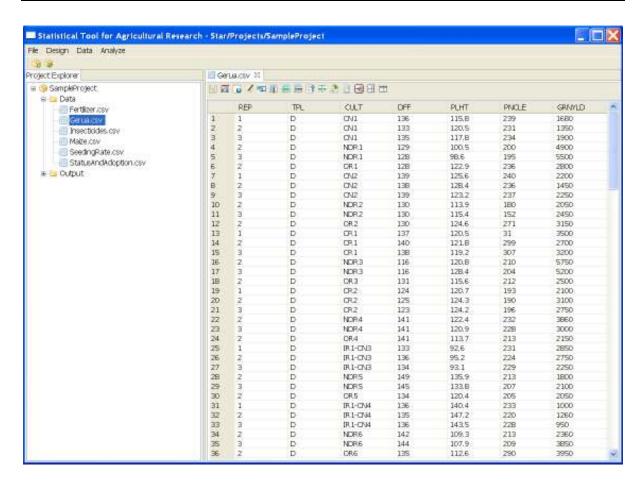


8.3. Split Plot Design in Randomized Complete Block

The steps to perform the Analysis of Variance using Split Plot Design in Randomized Complete Block are listed below:

• On the Project Explorer the dataset *Gerua.csv* located in the *Data* folder of the project named *SampleProject*. Double-click the file to open and view it in the Data Viewer. The example file contains data from an experiment conducted using Split Plot Design with transplanting method (*TPL*) as the main plot factor and cultivar (*CULT*) as the subplot factor in RCB. The experiment was done to compare the effects of two transplanting method (with levels *T* and *N*) using 25 cultivars with three blocks (*REP* with levels 1, 2, and 3) on grain yield (*GRNYLD*, in kg/ha) as well as other variables: plant height (*PLHT*, in cm) and panicle number (*PNCLE*).

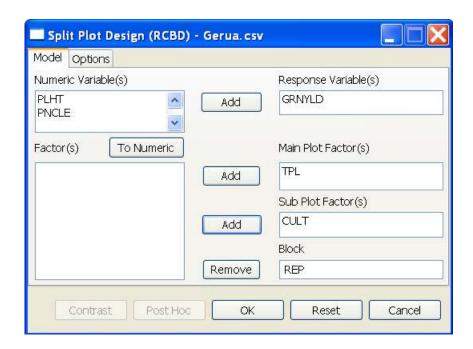
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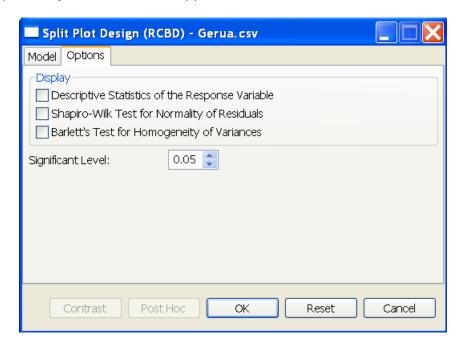
- From the main window of STAR click Analyze | Analysis of Variance | Split Plot Design | Randomized Complete Block Design (RCBD). The Split Plot Design (RCBD) dialog box will appear.
- Opening the data for the first time, REP field in the data file is regarded by R as numerical variable; it need to be changed as factors. Choose the variable and click on the To Factor button.
- Specify the required field and appropriate options for the analysis.

For the example, the completed **Model** Specification tab should appear as illustrated below:

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The completed **Options** tab should appear as illustrated below:



- Click OK button to perform the analysis. The Split Plot Design (RCBD) dialog box will be minimized and STAR activates the Output Viewer tab and the Graph Viewer tab.
- Sample text output of the analysis (only results from response variable = GRNYLD) displayed in the Output Viewer tab is shown below:

Analysis of Variance Split Plot Design in RCBD

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CLASS INFORMATION

Class Level Information

FACTOR NO. OF LEVELS LEVELS

REP 3 1, 2, 3 TPL 2 D, N

CULT 26 CN1, CN2, ..., S2

Number of Observations Read and Used: 156

ANOVA TABLE

Response Variable: GRNYLD

Source	DF	Sum of Square	Mean Square	F Value	Pr > F
REP TPL Error(a)	2 1 2	58473.3974 109947636.0577 168641.3462	29236.6987 109947636.0577 84320.6731	0.35 1303.92	0.7425
CULT TPL:CULT Error(b) Total	25 25 100 155	134144970.6731 16611234.7756 10517968.5897 271448924.8397	5365798.8269 664449.3910 105179.6859	51.02 6.32	0.0000

Summary Statistics

Coef Var(a) Coef Var(b) GRNYLD Mean
7.72 8.63 3759.20

Comparison of TPL at each level of CULT

Least Significant Difference (LSD) Test

Alpha	0.05
Error Degrees of Freedom	101
Error Mean Square	104377.4162
Critical Value	1.9837
Test Statistic	523.2683

Summary:

TPL	N	CULT = CN1 group	CULT = CN2 group	 CULT = S2 group
D N	3		1966.67 b 3763.33 a	 2163.33 b 4260.00 a

^{*} Means with the same letter are not significantly different

Comparison of CULT at each level of TPL

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Tukeys's Honest Significant Difference (HSD) Test

Alpha	0.05
Error Degrees of Freedom	100
Error Mean Square	105179.6859
Critical Value	5.3519
Test Statistic	1002.1128

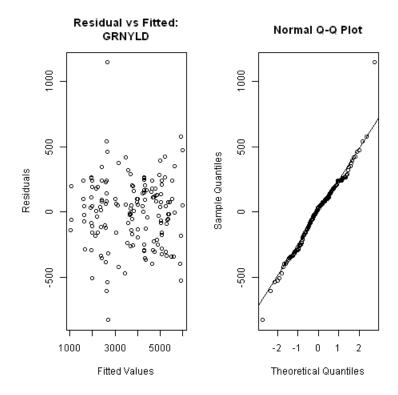
Summary:

CULT	N	TPL = D	group	TPL = N	group
CN1	3	1643.33	 j k	3683.33	 hij
CN2	3	1966.67	ijk	3763.33	
CR1	3	3133.33	efgh	5033.33	abcde
CR2	3	2650.00	ghi	4270.00	efghi
IR1-CN3	3	2616.67	ghij	5050.00	abcde
IR1-CN4	3	1070.00	k	3026.67	j
IR2	3	1950.00	ijk	4260.00	efghi
NDR1	3	5216.67	ab	5890.00	ab
NDR2	3	2416.67	hij	4926.67	bcdef
NDR3	3	5520.00	a	5956.67	a
NDR4	3	3453.33	defg	4690.00	cdefg
NDR5	3	1966.67	ijk	3986.67	fghij
NDR6	3	2686.67	fghi	4830.00	cdef
NDR7	3	4783.33	abc	5333.33	abcd
OR1	3	2583.33	ghij	5250.00	abcde
OR2	3	2616.67	ghij	4326.67	efghi
OR3	3	2416.67	hij	4400.00	defghi
OR4	3	1966.67	ijk	4000.00	fghij
OR5	3	2216.67	hij	4653.33	cdefgh
OR6	3	3800.00	cde	5033.33	abcde
OR7	3	5066.67	ab	5616.67	abc
OR8	3	4316.67	bcd	5386.67	abcd
RA1	3	1633.33	jk	3763.33	ghij
RA2	3	3673.33	def	4616.67	cdefgh
S1	3	2385.00	hij	3560.00	ij
S2	3	2163.33	hij	4260.00	efghi

^{*} Means with the same letter are not significantly different

Sample graphical output of the analysis displayed in the Graph Viewer tab is shown below:

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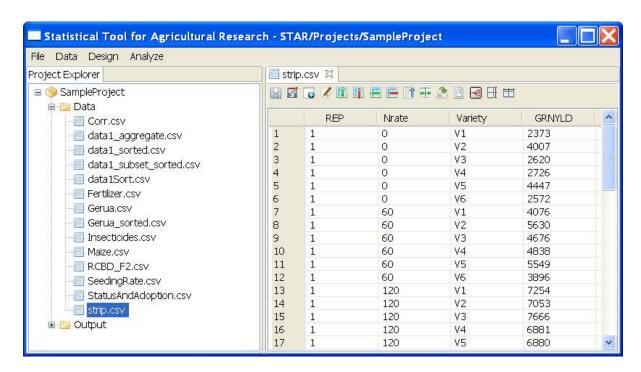


8.4. Strip Plot Design

The steps to perform the Analysis of Variance in Randomized Complete Block Design are listed below:

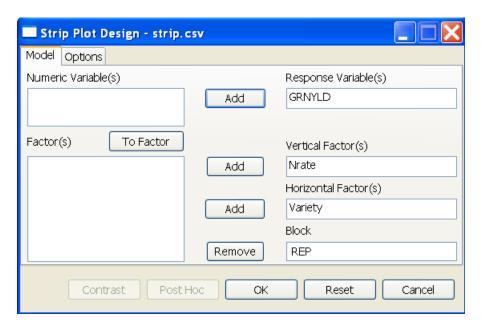
• On the Project Explorer the dataset *strip.csv* located in the *Data* folder of the project named *SampleProject*. Double-click the file to open and view it in the Data Viewer. The example file contains data from an experiment conducted to compare the effects of six rice variety (*Variety*, with levels *V1*, *V2*, *V3*, *V4*, *V5* and *V6*) and three rates of nitrogen (*Nrate*, in kg/ha with levels *0*, *60* and *120*) on grain yield (*GRNYLD*, in kg/ha). A strip plot design with four blocks (*REP* with levels *1*, *2*, and *3*) was used with rice variety as the horizontal factor and nitrogen rates as the vertical factor.

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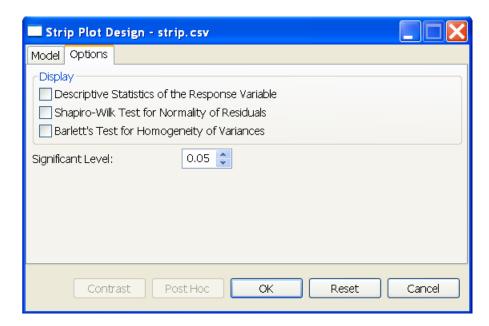


- From the main window of STAR click Analyze | Analysis of Variance | Strip Plot Design. The Strip Plot dialog box will appear.
- Opening the data for the first time, REP and Nrate fields in the data file are regarded by R as numerical variables; they need to be changed as factors. Choose the variable and click on the To Factor button.
- Specify the required field and appropriate options for the analysis.

For the example, the completed **Model** Specification tab should appear as illustrated below:



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The completed **Options** tab should appear as illustrated below:

- Click OK button to perform the analysis. The Strip Plot Design dialog box will be minimized and STAR activates the Output Viewer tab and the Graph Viewer tab.
- Sample text output of the analysis displayed in the Output Viewer tab is shown below:

```
Analysis of Variance
Strip Plot Design
CLASS INFORMATION
Class Level Information
FACTOR
         NO. OF LEVELS LEVELS
______
REP
                       3 1, 2, 3
Nrate
                       3 0, 60, 120
                       6 V1, V2, ..., V6
Variety
Number of Observations Read and Used: 54
ANOVA TABLE
Response Variable: GRNYLD
           DF Sum of Square Mean Square F Value Pr > F
_____
REP
                        9215077.1481 4607538.5741
                                                          6.19 0.0596
Nrate 2 50672299.5926 25336149.7963 34.05 0.0031

Error(a) 4 2975911.9630 743977.9907

Variety 5 57101029.4259 11420205.8852 7.65 0.0034

Error(b) 10 14922379.9630 1492237.9963

Nrate:Variety 10 23874076.8519 2387407.6852 5.80 0.0004
```

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Error(c)	20	8230550.9259	411527.5463
Total	53	166991325.8704	

Summary Statistics

Coef Var(a)	Coef Var(b)	Coef Var(c)	GRNYLD Mean
16.31	23.09	12.13	5289.76

Comparison of Nrate at each level of Variety

Least Significant Difference (LSD) Test

Alpha	0.05
Error Degrees of Freedom	22
Error Mean Square	466935.9537
Critical Value	2.0716
Test Statistic	1155.8252

Summary:

Nrate N	Vari	ety =	V1	group	Variety =	V2	group	 Variety = V6	group
0 3 120 3 60 3	3	3571. 7548. 5132.	00	a	4934 7211 6713	.33	a	 3207.33 2492.00 3714.33	b

^{*} Means with the same letter are not significantly different

Comparison of Variety at each level of Nrate

Tukeys's Honest Significant Difference (HSD) Test

Alpha	0.05
Error Degrees of Freedom	10
Error Mean Square	1492237.9963
Critical Value	4.9120
Test Statistic	3464.3214

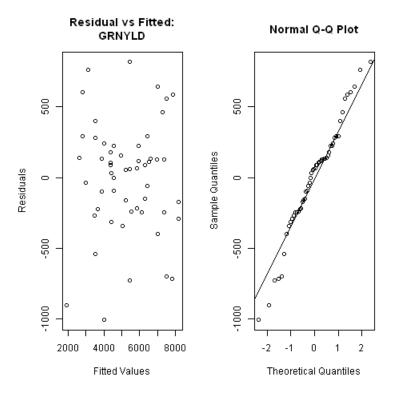
Summary:

Variety	N	Nrate = 0 group	Nrate = 60 group	Nrate = 120 group
V2 V3 V4 V5	3 3 3 3 3	3571.67 a 4934.33 a 4249.67 a 4059.00 a 4101.67 a	5132.00 a 6713.67 a 6122.33 a 5553.67 a 5630.00 a	7548.00 a 7211.33 a 7868.33 a 7094.33 a 6012.00 a
V6	3	3207.33 a	3714.33 a	2492.00 b

^{*} Means with the same letter are not significantly different

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Sample graphical output of the analysis displayed in the Graph Viewer tab is shown below:



8.5. Partitioning Sum of Squares

The **Contrast** option is used in partitioning sum of squares that decompose the variability of the response(s) into various components to have a better view of the nature of variability. There are different selections in partitioning sum of squares such as, "Compare with Control", "User Specified Contrast (Group Comparison)" and "Orthogonal Polynomial Contrast".

8.5.1. Group Comparison

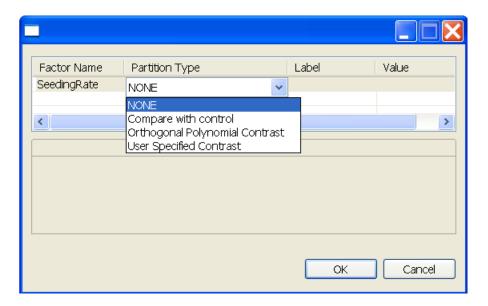
The steps to perform the Group Comparison are listed below:

- On the Project Explorer the dataset *SeedingRate.csv* located in the *Data* folder of the project named *SampleProject*. Double-click the file to open and view it in the Data Viewer.
- Perform the Analysis of Variance. Follow the discussion in Randomized Complete Block Design (RCBD) of this module.
- After performing the Analysis of Variance, the Contrast button will be enabled if there is at least one factor in the analysis with more than two levels.

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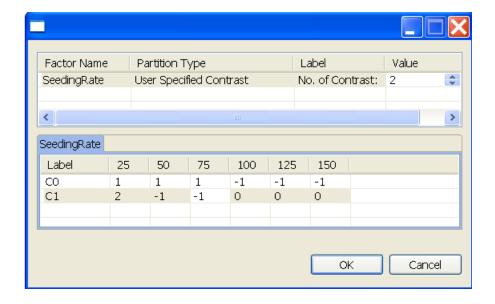
Suppose we want to determine the following:

- i. If the mean of the SeedingRate 25, 50, and 75 is significantly different from SeedingRate 100, 125 and 150
- ii. If the mean of the SeedingRate 25, is significantly different from SeedingRate 50, and 75
- Click the Contrast button. A dialog box will appear.



Specify the required fields and appropriate options for the analysis. In the **Partition** Type column, click the drop-down menu and choose *User Specified Contrast*. Specify
 the number of contrast. A tab will be displayed in the lower part of the Constrast Analysis
 dialog box.

For the example, the completed dialog box should appear as illustrated below:



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- Click the **OK** button to perform Group Comparison. The Contrast Analysis dialog box will be closed and STAR actives the Output Viewer tab.
- Sample output of the analysis:

8.5.2. Trend Comparison

This procedure is applicable and available if the levels of the factor are quantitative.

The steps to perform the Group Comparison are listed below:

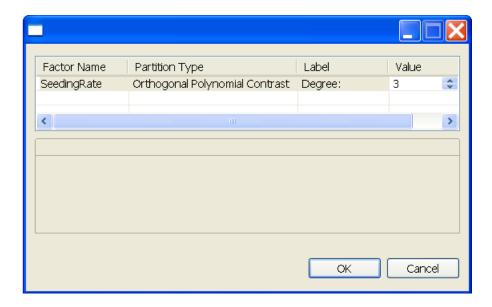
- On the Project Explorer the dataset *SeedingRate.csv* located in the *Data* folder of the project named *SampleProject*. Double-click the file to open and view it in the Data Viewer.
- Perform the Analysis of Variance. Follow the discussion in Randomized Complete Block Design (RCBD) of this module.
- After performing the Analysis of Variance, the Contrast button will be enabled if there is at least one factor in the analysis with more than two levels.

Suppose we also want to know the degree of the relationship between yield response and Seeding Rate.

- Click the Contrast button. A dialog box will appear.
- Specify the required fields and appropriate options for the analysis. In the **Partition Type** column, click the drop-down menu and choose *Orthogonal Polynomial Constrast*.

 Specify the highest degree polynomial that will be fitted.

For the example, the completed dialog box should appear as illustrated below:



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- Click the **OK** button to perform Trend Comparison. The Contrast Analysis dialog box will be closed and STAR actives the Output Viewer tab.
- Sample output of the analysis:

Result of Contrast Analysis

ANOVA TABLE

Response Variable: GrainYield

Source	 DF	Sum of Square	Mean Square	F Value	Pr > F
Rep SeedingRate SeedingRate:	3 5 1	1.7667 10.1800 6.7891 1.6296	0.5889 2.0360 6.7891 1.6296	4.87 16.84 56.16 13.48	0.0147 0.0000 0.0000 0.0023
SeedingRate: SeedingRate: Error Total	1 15 23	1.6296 0.1027 1.8133 13.7600	0.1027 0.1209	0.85	0.0023

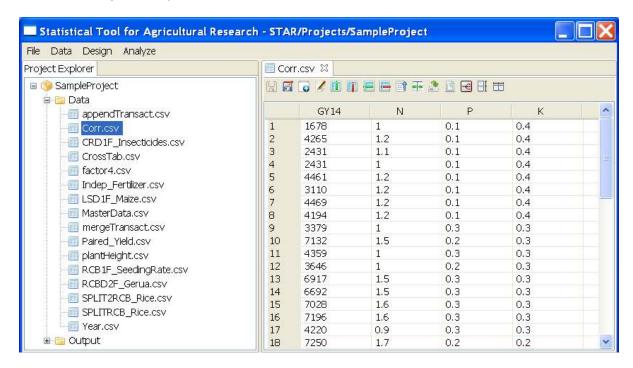
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9. Correlation Analysis

The **Correlation Analysis** is used to determine the strength of linear relationship between variables. The **Correlation Analysis** submenu performs bivariate correlation using the pearson's correlation coefficient, spearman's rho and kendall tau. By default, it uses the pearson's correlation coefficient.

The steps to perform Correlation Analysis are listed below:

On the Project Explorer, locate the dataset Corr.csv in the Data folder of the project named SampleProject. Double-click the file to open and view it in the Data Viewer tab. The example file contains data from an experiment conducted to investigate the relationship between grain yield (GY14, in kg/ha), the grain's nitrogen (N), phosphorus (P) and potassium (K) content in percent. The data were collected from 48 experimental plots with grains adjusted to 14% moisture content.



- Click Analyze | Correlation Analysis from the main window of STAR. The Correlation Analysis dialog box will appear.
- Specify the required field and appropriate options for the analysis:

Variable Description Tab

Test Variable(s)

This field is required. For the analysis to proceed, this list box should have at least two entry.

Correlation Coefficient

For the analysis to proceed, at least one of the methods should be tick. By default, Pearson's correlation coefficient is computed. For quantitative and normally distributed variables, choose the Pearson correlation coefficient. If the data are not normally distributed or ordinal level, choose Kendall's tau or Spearman, which measures the association between ranks. The correlation coefficient ranges from -1 to +1.

Option Tab

Alternative Hypothesis

User can select a one-tailed or two-tailed test. The default alternative hypothesis used is that for the two-sided test.

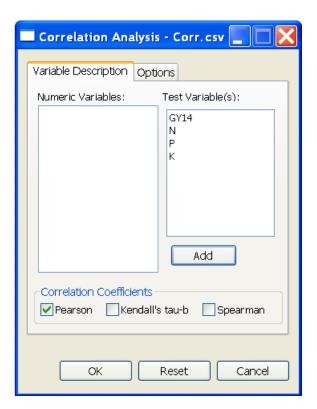
Display Summary Statistics

If this option is selected, a summary table with number of observations, minimum, maximum, mean and standard deviation will be displayed.

<u>Scatterplot</u>

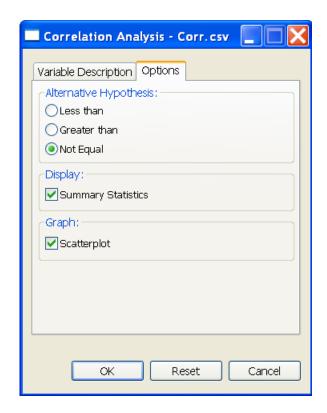
If this option is selected, scatter plot for all pair of variables listed in the **Test Variable(s)** list box will be created.

For the example, the completed **Variable Description** tab should appear as illustrated below:



The completed **Option** tab should appear as illustrated below:

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- Once all options have been specified, click the OK button. The Correlation Analysis
 dialog box will be minimized and STAR activates the Output Viewer which shows the
 results of the analysis.
- Sample output of the Correlation Analysis using Pearson's product-moment correlation is shown below.

Result of Correlation Analysis

	~
DESCRIPTIVE	STATISTICS

Variable	N_NonMissObs	Min	 Max	Mean	StdDev
GY14 N P K	48 48 48	1678.00 0.10 0.10 0.20	7860.00 1.80 0.30 0.40	4788.67 1.23 0.23 0.31	1552.27 0.33 0.08 0.06

CORRELATION ANALYSIS

Pearson's product-moment correlation, Prob > |r|

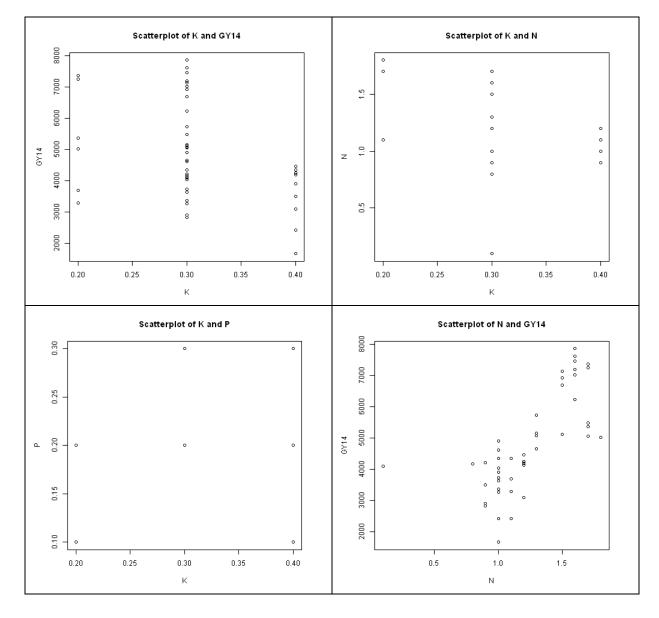
		GY14	N	P	K
GY14	coef	1.00	0.72	0.38	-0.40
	p-value		0.0000	0.0080	0.0053
	n	48	48	48	48
N	coef	0.72	1.00	0.02	-0.34
	p-value	0.0000		0.8686	0.0180
	n	48	48	48	48

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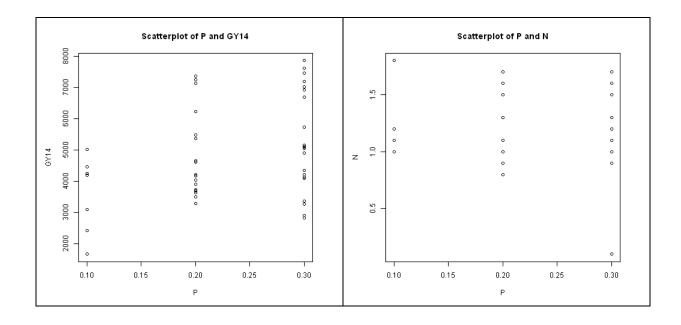
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P	coef	0.38	0.02	1.00	-0.35
	p-value	0.0080	0.8686		0.0161
	n	48	48	48	48
K	coef	-0.40	-0.34	-0.35	1.00
	p-value	0.0053	0.0180	0.0161	
	n	48	48	48	48

• If a scatter plot is requested, the graph can be viewed in the **Graph Viewer**. Sample generated graphs are shown below:



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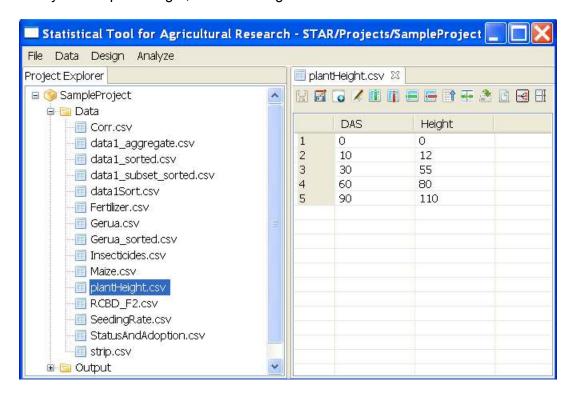
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10. Linear Regression Analysis

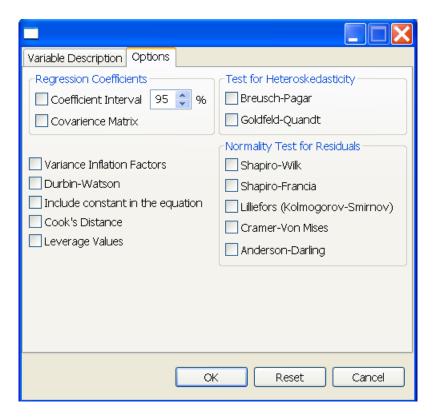
Regression Analysis is the statistical technique used to determine the strength of linear relationship between variables.

The steps to perform Linear Regression Analysis are listed below:

On the Project Explorer, locate the dataset *plantHeight.csv* in the *Data* folder of the project named *SampleProject*. Double-click the file to open and view it in the Data Viewer tab. The file contains the data of number of days after seeding (DAS) as independent variable and plant height as dependent variable. Based on this data, simple linear regression analysis is to be conducted to assess how the days after seeding linearly affects plant height, on the average.



Click Analyze | Regression Analysis from the main window of STAR. The Regression dialog box will appear.



Specify the required field and appropriate options for the analysis:

Variable Description Tab

Dependent Variable

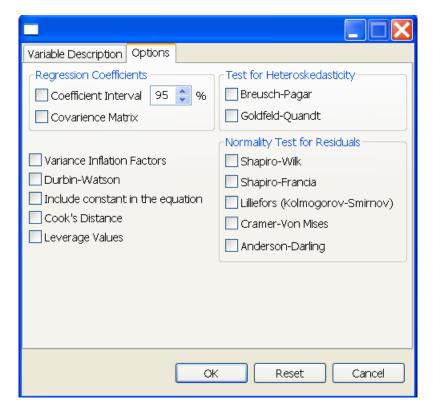
This field is required. For the analysis to proceed, this list box should have at least one entry.

Independent Variables

This field is required. For the analysis to proceed, this list box should have at least one entry.

Option Tab

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Regression Coefficients

User can choose to display the confidence interval and the covariance matrix. If any of this is selected, the results is embedded in the table of parameter estimates. For the confidence interval, the default value is 95%. Valid value is between 90% to 99%.

Test for Heteroskedasticity

There are two available procedures namely: Breusch-Pagar and Goldfeld-Quandt.

Normality Test

If this option is selected, it will perform normality test for the residuals. There are five procedures available, namely: Shapiro-Wilk, Shapiro-Francia, Lilliefors, Cramer-Von Mises and Anderson-Darling.

Variance Inflation Factors

If this is selected, it displays the variance inflation factors (VIF) which measures the Collinearity (or multicollinearity).

Durbin-Watson

If this is selected, it displays the Durbin-Watson test for serial correlation of the residuals.

Include constant in the equation

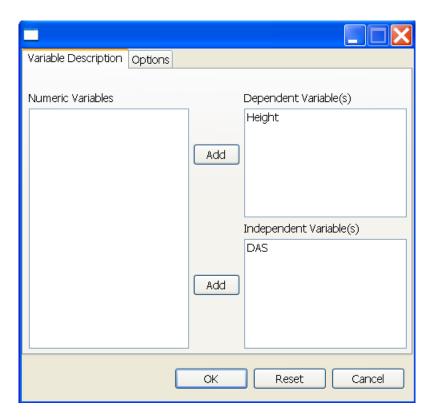
By default, the regression model includes a constant term in the equation. Deselecting this option forces regression through the origin, this is rarely done.

Cook's Distance

Leverage Values

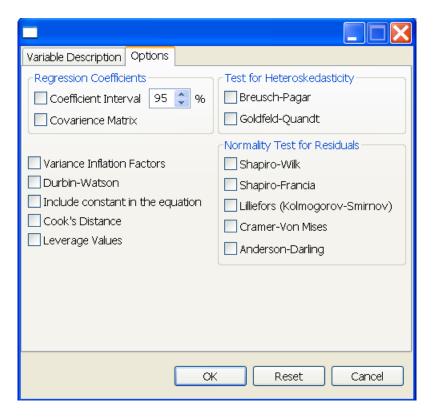
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For the example, the completed **Variable Description** tab should appear as illustrated below:



The completed **Option** tab should appear as illustrated below:

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Once all options have been specified, click the OK button. The Regression Analysis
dialog box will be minimized and STAR activates the Output Viewer which shows the
results of the analysis.

Output Viewer

The default output of the Regression Analysis procedure in STAR includes the Analysis of Variance table, a table of model summary and the parameter estimated. The output starts with the descriptive statistics, if this option was specified in the **Options** tab.

Graph Viewer

The tab displays two diagnostics plots. The left plot, entitled *Residuals vs. Fitted* is for assessing the distribution of the residuals. An ideal plot should reveal that the points are distributed around 0. The residuals should not be affected by the size of the fitted values. Problematic data may result to a plot wherein the points increase as the fitted values increase, thus forming a "funnel-shaped" distribution. This will indicate a violation of the homogeneity of variances requirement for analysis of variance. The *Normal Q-Q* plot is used for assessing normality. The points should lie about the line.

Sample output of the Regression Analysis is shown below.

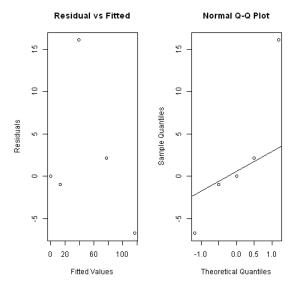
DESCRIPTI	VE STATISTICS			
Variable	N_NonMissObs	Mean	StdDev	SE_Mean
Height	 5	51.40	45.99	20.57

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DAS			5	38.00	37.0	1	16.55			
LINEAR R Model Fi	tted:	Height	. ~ DA	AS						
Analysis	oi Va									
Source		Sum of								
Model DAS Error Total	1 1 3	820 820 25)1.389)1.389 57.810	98 98 02	8201.3 8201.3	898 898				
Model Su		:								
Root MS	E He	ight Mea	an Co	oeff V	ar R-	Squar		j R-S	- 1	
	7	51.4	10	89.	47	0.969	95 (4	
Paramete	r Est:	imates:							-	
Variable	E:	stimate	Std	. Erro	or t v	alue	LL C	I* UI	L CI*	Pr(> t)
Intercep DAS	t	4.91 1.22		6.3 0.1	31	0.78 9.77	-15.1 0.8	17 2 82	25.00	0.4931
* At 95%										
TEST FOR	-	ALITY								
Variable	Metl		St	tat	Value					
residual		piro-Wil	.k W							
Durbin W lag Aut										

The graph can be viewed in the **Graph Viewer**. Sample generated graphs are shown below:

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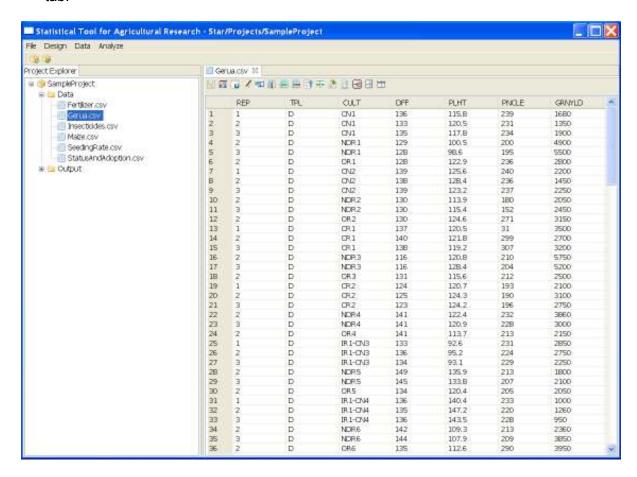
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11. Normality Test

The Test for Normality menu display normality test for at least one numeric variable. There are five procedures available in this menu, namely: Shapiro-Wilk, Shapiro-Francia, Lilliefors (Kolmogorov-Smirnov), Cramer-Von Mises and the Anderson-Darling test for normality.

The steps to perform Normality Test are listed below:

 On the Project Explorer, locate the dataset Gerua.csv from the Data folder of the project named SampleProject in. Double-click the file to open and view it in the Data Viewer tab



- Click Analyze | Descriptive Statistics | Normality Test. The Normality Test dialog box will appear.
- Opening the data for the first time, Rep field is regarded by R as numerical variables, they need to be changed as factors. Choose the variable and click on the To Factor button.
- Specify the required field and appropriate options for the analysis.

Variable Description Tab

Variables

This field is required. At least one item should be specified for the analysis to proceed.

By Variables

This field is optional. If a **By Variable(s)** is specified, it will perform test for normality per level of the **By Variable(s)**.

Options Tab

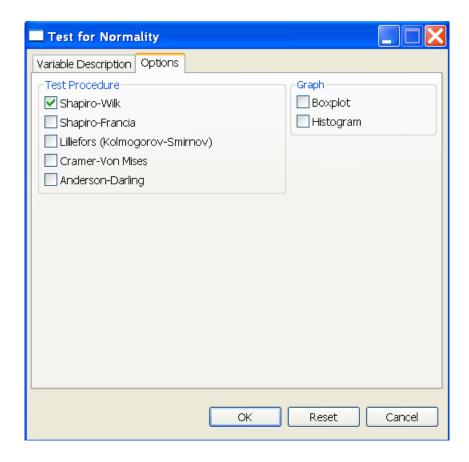
Test Procedure

This option is required. At least one test procedure should be tick for the analysis to proceed. By default, the Shapiro-Wilk test is performed.

Graph

The user has the option to display the box plot and the histogram.

For the example, the complete dialog box should appear as shown below:



 Click Ok button to perform the test. The Normality Test dialog box will be minimized and STAR activates the Output Viewer tab.

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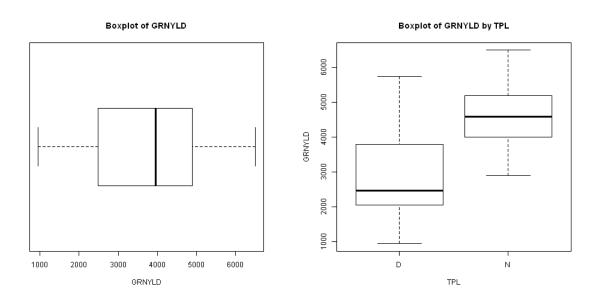
Sample output of the normality test for variable GRNYLD.

TEST FOR	NORMALITY			
Variable	Method	Stat	Value	p Value
GRNYLD PNCLE PLHT DFF	Shapiro-Wilk Shapiro-Wilk Shapiro-Wilk Shapiro-Wilk	พ พ พ	0.9624 0.9294 0.9580 0.8137	0.0003 0.0000 0.0001 0.0000

Sample output of the normality test for variable GRNYLD by TPL.

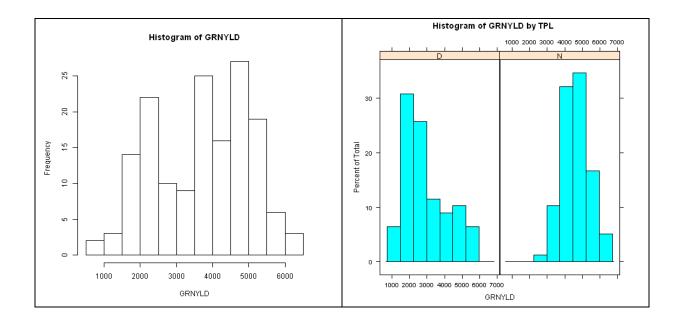
TEST	FOR NO	RMALITY				
Grp	Level	Variable	Method	Stat	Value	p Value
TPL TPL	D N	GRNYLD GRNYLD	Shapiro-Wilk Shapiro-Wilk	W W	0.9176 0.9908	0.0001 0.8517

• If at least one of the graphs is chosen it can be viewed in the **Graph Viewer** tab. Below is the sample output if the boxplot was chosen:



Below is the sample output if the histogram was chosen:

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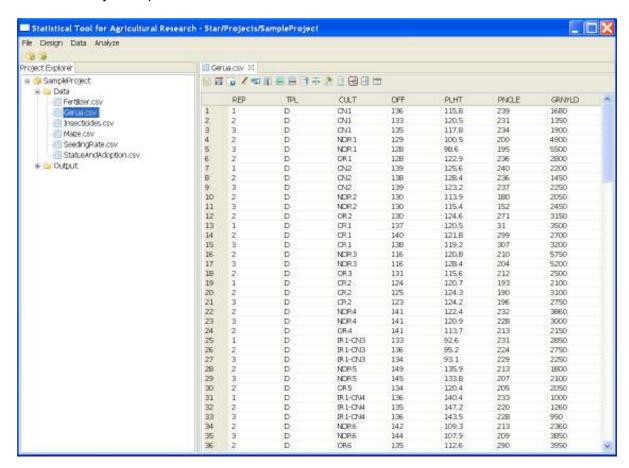
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12. Heterogeneity Test

The Heterogeneity Test menu perform test for equally of variances. There are two procedures available in this menu, namely: Bartlett's test and the Levene's test.

The steps to perform Heterogeneity test are listed below:

• Locate the dataset *Gerua.csv* from the *Data* folder of the project named *SampleProject* in the Project Explorer.



- Click Analyze | Descriptive Statistics | Heterogeneity Test.
 The Heterogeneity Test dialog box will appear.
- Opening the data for the first time, Rep field is regarded by R as numerical variables, they need to be changed as factors. Choose the variable and click on the To Factor button.
- Specify the required field and appropriate options for the analysis.

Variable Description Tab

<u>Variables</u>

This field is required. At least one item should be specified for the analysis to proceed.

By Variables

This field is required. All items specified here should contain two or more levels.

Options Tab

Test Procedure

This option is required. At least one test procedure should be tick for the analysis to proceed. By default, the Bartlett's test for homogeneity of variances is performed.

<u>Graph</u>

The user has the option to display the box plot and the histogram.

For the example, the complete dialog box should appear as shown below:

Suppose we want to determine whether the variances are equal among the levels of *TPL* for variables *DFF*, *PLHT*, *PNCLE* and *GRNYLD*. For the *example*, the completed **Variable Description** tab should appear as illustrated below:

The completed **Option** tab should appear as illustrated below:

- Click Ok button to perform the test. The Normality Test dialog box will be minimized and STAR activates the Output Viewer tab.
- Sample output of the normality test for variable GRNYLD.

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13. Nonparametric Tests

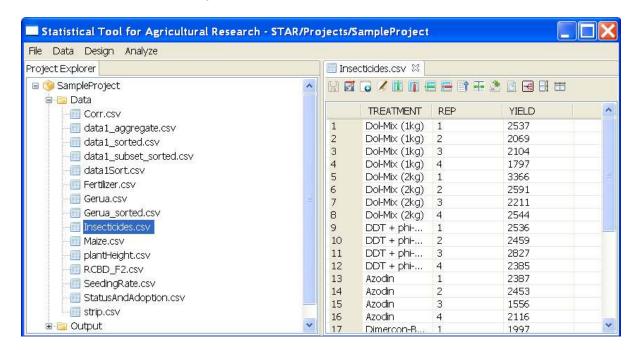
Nonparametric Tests procedure provides several tests that do not require assumptions about the shape of the underlying distribution.

13.1. Test for Several Independent Samples

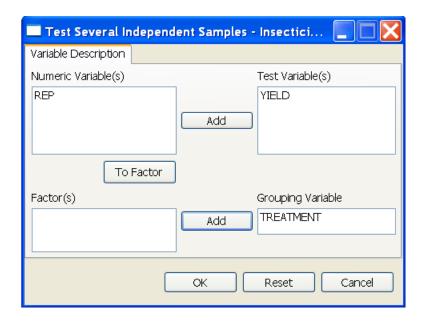
This procedure perform Kruskal Wallis test.

The steps to perform Kruskal Wallis test are listed below:

• Locate the dataset *Insecticides.csv* from the *Data* folder of the project named *SampleProject* in the Project Explorer.



- Click Analyze | Non-Parametric Test | Test for Several Independent Samples. The Test for Several Independent Samples dialog box will appear.
- Specify the required field and appropriate options for the analysis. The completed dialog box should appear as shown below:



• Sample output is shown below:

Rank Sums for Variable YIELD Classified by Variable TREATMENT

TREATMENT	N	Sum of	Ranks	Mean of	Ranks
Azodin Control	 4 4		64 12		16.00
DDT + phi-BHC	4		91		22.75
Dimercon-Boom	4		41		10.25
Dimercon-Knap	4		35		8.75
Dol-Mix (1kg)	4		66		16.50
Dol-Mix (2kg)	4		97		24.25

^{*} Average scores are used for ties.

Kruskal-Wallis	rank	sum	test
Chi-Square		20	.8522
DF			6
Pr > Chi-Square	9	(0.0020

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14. Appendix

14.1. Data Description

14.1.1. Insecticides.csv

Description

Grain yield of rice resulting from use of different foliar and granular insecticides for the control of brown planthoppers and stem borers from a Completely Randomized Design Experiment with Four Replicates and 7 treatments. The data contains 28 rows and 3 columns.

Formats

This data contains the following columns:

Treatment a factor with 7 levels (Azodin, DDT + γ -BHC, Dimercon-Boom,

Dimercon-Knap, Dol-Mix (1 kg), Dol-Mix (2 kg) and Control) denoting

the Different Foliar and Granular Insecticides

Rep a factor with 4 levels (1 to 4) denoting the replicates

GrainYield a numeric vector of grain yield in kg/ha

Source

Gomez, Kwanchai A. and Arturo A. Gomez. (1984) *Statistical Procedures for Agricultural Research 2nd Edition*.

14.1.2. Maize.csv

Description

Grain yield of three promising maize hybrids and a check variety from an experiment using Latin Square design. The data contains 24 rows and 3 columns.

Formats

This data contains the following columns:

Maize a factor with 4 levels denoting the Maize Hybrids (A, B, and D) and a

Check Variety (C)

Row a factor with 4 levels (1 to 4) denoting the row blocking factor

Column a factor with 4 levels (1 to 4) denoting the column blocking factor

GrainYield a numeric vector of grain yield in kg/ha

Source

Gomez, Kwanchai A. and Arturo A. Gomez. (1984) *Statistical Procedures for Agricultural Research 2nd Edition*.

14.1.3. SeedingRate.csv

Description

Grain Yield of Rice Variety IR8 with Six Different Rates of Seeding, from a Randomized Complete Block Design Experiment with Four Replicates. The data contains 24 rows and 3 columns.

Formats

This data contains the following columns:

Seeding Rate in kg seed/ha with 6 levels (25, 50, 75, 100, 125 and

150) denoting the treatment

Rep block with 4 levels (1 to 4)

GrainYield a numeric vector of grain yield in kg/ha

Source

Gomez, Kwanchai A. and Arturo A. Gomez. (1984) *Statistical Procedures for Agricultural Research 2nd Edition*.

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