# **DSX Local SPSS Modeler Overview**

#### Overview

In this lab you will learn how to implement analytics in **SPSS Modeler**, a well-known visual data mining workbench which can be used in the **Data Science Experience** (DSX). This lab will introduce the SPSS Modeler capability using the Titanic dataset. The lab will guide the development of an SPSS Modeler stream that will prepare the input data for modeling to run a machine learning algorithm predicting survivability of a passenger on the Titanic. You can learn more about Modeler in DSX in official product documentation: <a href="https://content-dsxlocal.mybluemix.net/docs/content/local-dev/spss-modeler.html">https://content-dsxlocal.mybluemix.net/docs/content/local-dev/spss-modeler.html</a>

#### Introduction

**SPSS Modeler** is a visual data mining workbench. Modeler can be used to complete all tasks of the analytic application development

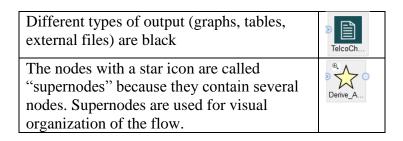
- Data understanding
- Data preparation
- Model building
- Model evaluation

Deployment is done in **Watson Machine Learning**. We discuss deployment in more detail in the **Watson Machine Learning** hands-on lab.

Assets developed in Modeler are called "flows". Another frequently used term in Modeler documentation is "streams" (used in Modeler desktop documentation). A flow starts with one or several data sources. Using visual nodes, a user can apply different operations to data. Data "flows" from one node to another in the direction of the arrows.

Visual nodes in modeler are color-coded and organized by type of operation: **Record Operations, Field Operations, Graphs, Modeling, Output,** and **Export** (data sources). Most operations are well-known functions in data preparation and analytics, such as sampling, filtering, binning, etc.

The data sources are purple	custome
Data preparation operations are blue	
Algorithms are green	≥ → Q CHURN
The models that are created based on algorithms are orange	S CHURN

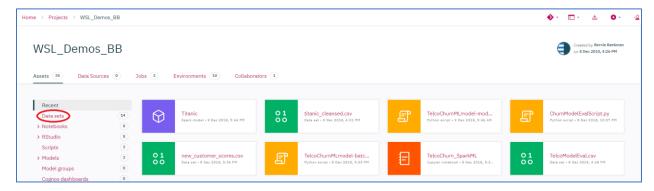


If a user needs more information about a particular node, it can be looked up in Modeler documentation. SPSS also publishes the **Algorithms Guide** that explains how machine learning algorithms are implemented in Modeler (see **Reference** for more information).

#### **Lab Steps**

## Step 1: Adding a Data Asset to the Watson Studio Local Labs project

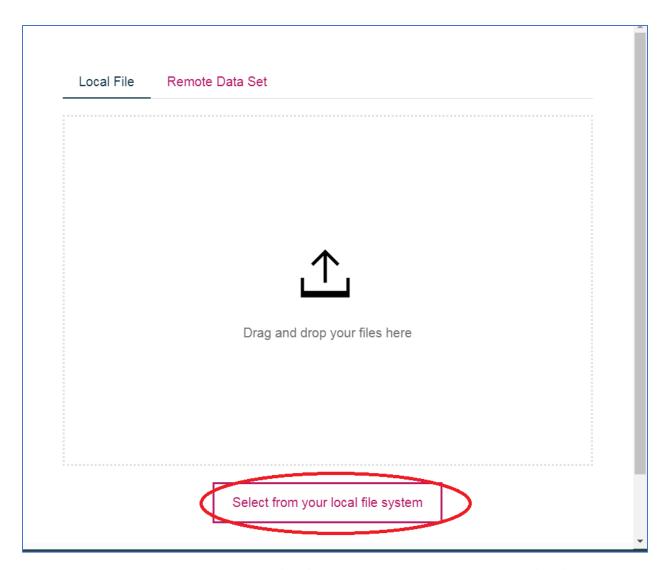
1. Click on **Data Sets** option in the Watson Studio Local project



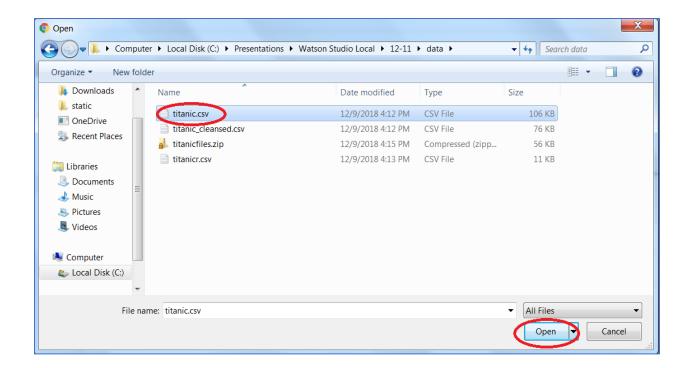
2. Click on Add Data Set.



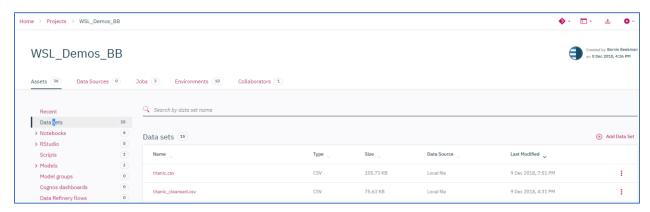
3. Click on **Select from your local file system**.



4. Navigate to the place where the **titanic.csv** was downloaded. Select the **titanic.csv** file and click **Open**.



5. The titanic.csv file is loaded into the project data sets.



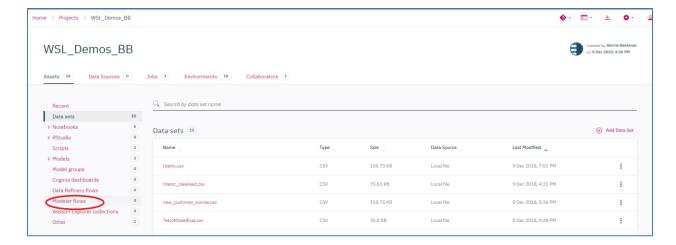
# Step 2: Create a Model to predict survival

In this section, we will create a Machine Learning flow using SPSS nodes. Documentation describing the nodes is available at <a href="https://content-dsxlocal.mybluemix.net/docs/content/local-dev/spss-modeler.html">https://content-dsxlocal.mybluemix.net/docs/content/local-dev/spss-modeler.html</a>.

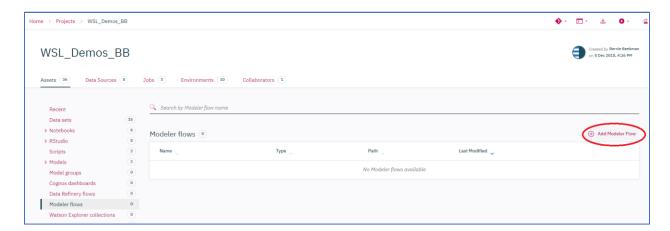
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# **Step 2.1 Create a New Flow and Load the Data**

1. In the DSX Local project, click on **Modeler flows** 



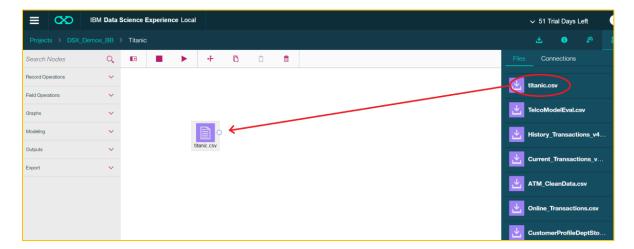
2. Click on Add Modeler Flow.



3. Click on the **Blank** tab, enter Titanic for the **Name**, optionally enter a **Description**, and click **Create**.



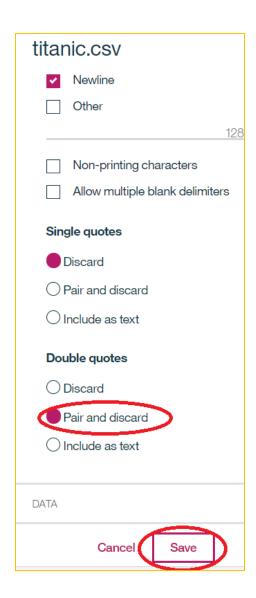
4. This opens the Flow Editor. Click on the hold down the left mouse key to drag it onto the canvas. Release the left mouse key.



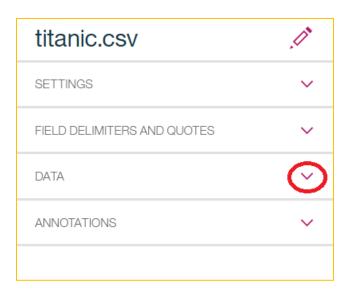
5. Double click on the titanic.csv icon. This will bring up the settings panel on the right-hand side. Click on the down arrow next to **FIELD DELIMITERS AND QUOTES**.



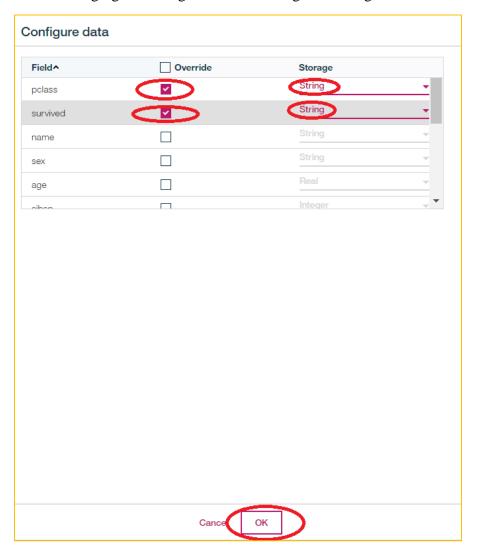
6. Scroll down and click on **Pair and discard** (if not already chosen) under Double quotes, and then click **Save**.



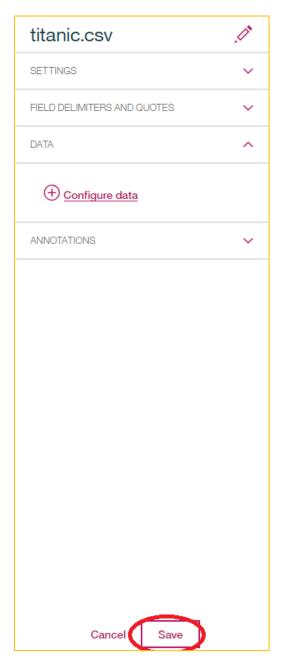
7. Double Click on the **titanic.cs**v icon again. This time select the down arrow next to **Data**, and click on **Configure data**.



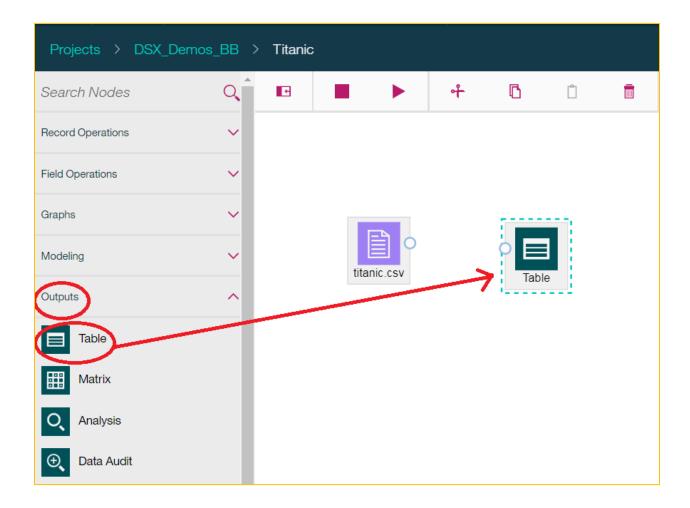
8. Override the pclass and survived storage classes, by clicking in the check box and changing the storage class from Integer to String, and then click **OK**.



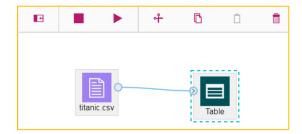
### 9. Click on Save.



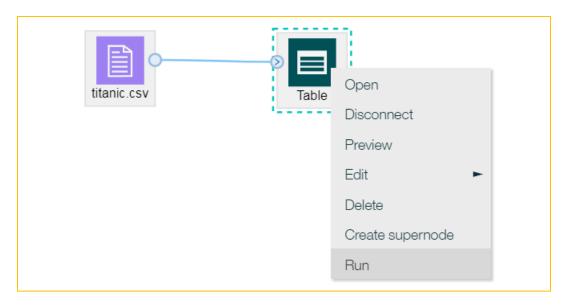
10. Click on the **Outputs** menu item in the Node Palette on the left and then click on the **Table** icon and drag the icon to the right of the titanic.csv icon. The SPSS Table node will display the contents of the csv file. If the Node Palette is not visible, click on the Node Palette icon



11. Connect the right side of the titanic.csv icon to the left side of the Table icon. This is accomplished by clicking on the little circle at the right side of the titanic.csv icon holding the left mouse key and dragging the mouse to the little circle on the left side of the Table icon, and then releasing the left mouse key.



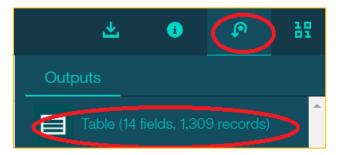
12. Right click on the **Table** icon and select **Run**.



13. The "Running Flow" prompt may appear and when completed a Table output selection will appear on the right side of the screen under the **Outputs** tab. If the Table output

selection does not appear, select the icon.





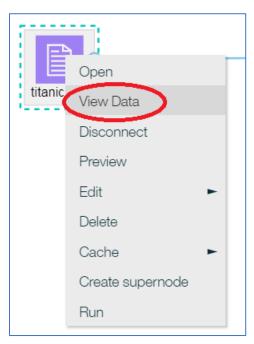
14. Double click on the Table selection and the contents of the titanic.csv will be displayed. Each row contains information on a passenger on the Titanic. We will use this data to make predictions on survivability.



### Step 2.2 Explore the Data using the View Data – Data Audit option.

Perusing through the data in the table, we can see that there are missing values. The SPSS Modeler has a View Data option that provides multiple ways to explore the data. The View Data option combines the capabilities of several SPSS modeler nodes (Table node, Data Audit node, and Visualization nodes) in one place.

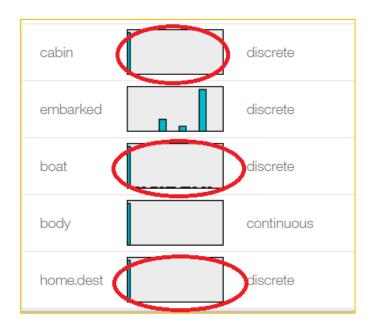




2. The **View Data** panel is shown below. Note, the screenshot below shows the **Spreadsheet** option. Four options appear at the left. The Spreadsheet option is similar to the Table node, the Data Audit option is similar to the Data Audit node, and the Chart option provides similar capabilities to the Graph nodes. The Preferences option provides different charting themes. Click on the **Data Audit** option.



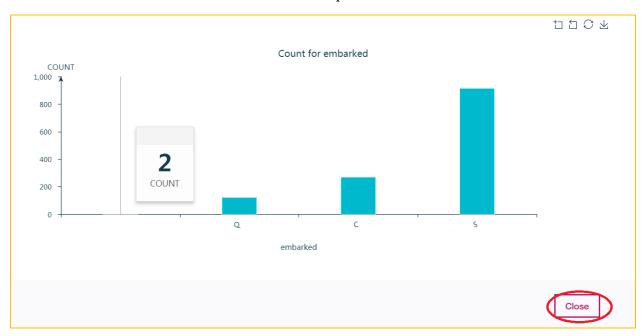
3. Data Audit provides profiling information on the input data that is useful for cleansing the data. It provides a comprehensive first look at the data, including summary statistics, as well as information about outliers, missing values, and extremes. You can click on the thumbnail images to display a larger image of the thumbnail with values. Click on the cabin, boat, and home dest thumbnails. We can see that each of these contains a significant number of blank or empty string values. These fields will be removed using a **Filter** node below.



4. Clicking on the cabin thumbnail displays a larger image. Hovering the cursor over the spike in the graph, we can see 1014 values out of 1309 are empty or blank strings. Click on **Close** to return to the Audit view.



**5.** Clicking on the embarked thumbnail icon shows that there are two missing values for this field. Click on **Close** to return to the **Audit** panel.



6. Click on the **Quality** tab. The Quality tab provides information on null values for continuous variables and displays outlier and extreme values. Note that quality information on discrete (string) fields is missing from this panel (this information does appear in the Quality information provided by the Data Audit node, so you may consider adding a Data Audit node as well). We see that there are quite a few missing values from the age field, 1 missing value from the fare field, and many missing values in the body field. We will remove the rows containing the missing values from the fare, embarked, and age fields using a **Select** node below. We will drop the body field in addition to the other fields mentioned above using a **Filter** node.

# Step 2.3 Explore the Data using the View Data – Chart Option

Let's explore the characteristics of the data using the Chart option.

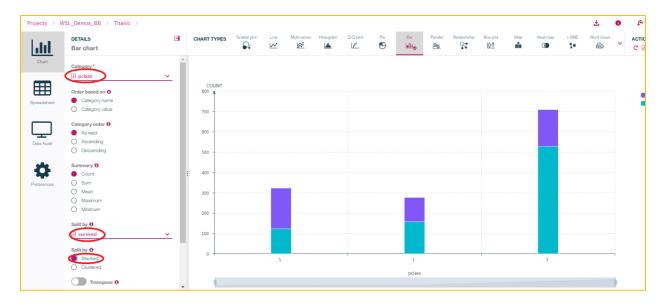
1. Click on the **Chart** option.



2. Click on Bar.



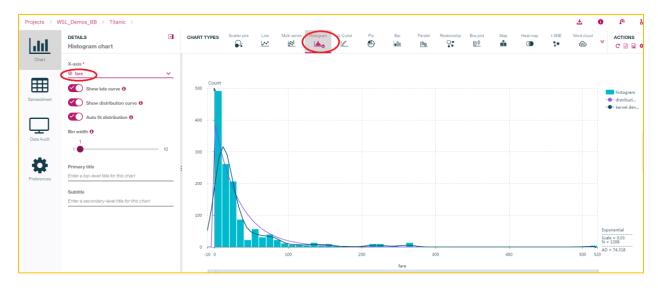
3. Select pclass for the **Category**, select survived for the **Split By**, and click on **Stacked**. We can see from the graph that the likelihood of surviving is correlated to the passenger class. The first-class passengers have the highest rate of survivability.



4. Change the **Category** to sex. The survivability rate is strongly influenced by the passenger sex.



5. Let's examine the histogram of the fare field. Click on **Histogram**, and select fare in the **Value** field We can see that the histogram is highly skewed. Skewness will impact the effectiveness of some machine learning techniques. One way to deal with skewness is to do a logarithmic transformation of the data. We will do this transformation in the preparing the data for modeling section below.



6. Note the above visualizations can also be done using **Graph** nodes. The **Distribution** node can be used to generate the bar charts above, and the **Histogram** node can be used to generate the histograms. The View Data option is very convenient in that these visualizations can be done in one place.

### Step 2.4 Prepare the Data for Modeling

Based on our exploration of the data, there are several transformations that are needed to prepare the data for modeling. This section will introduce, the **Filter** node, the **Select** node, and the **Derive** node that will do the necessary transformations. The **Filter** and **Derive** nodes act on a field level, whereas the **Select** node acts on a record level.

**Filter** node – The **Filter** node performs two functions. It specifies fields that can be dropped. It also allows fields to be renamed. We will drop the fields cabin, boat, body, and home.dest.

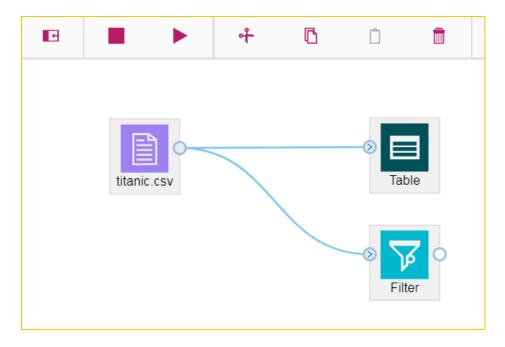
**Derive** node — The **Derive** node modifies data values or creates new fields from one or more existing fields. We will use the derive node to do a logarithmic transformation of the fare field. We will also use this node to bin the age and fare fields.

**Select** node – The **Select** node is used to select or discard a subset of records from the data stream based on a specific condition. We will remove the rows where there is missing information in the fare, age, or embarked fields.

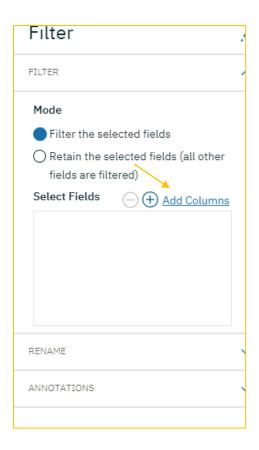
1. Click **Titanic** to return to the canvas.



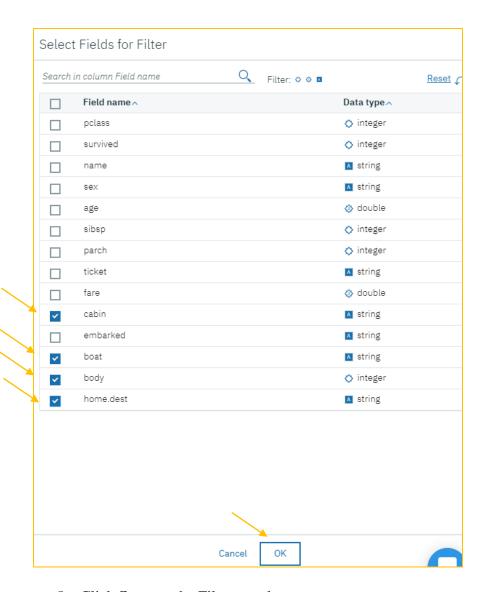
2. Add a **Filter** node to drop fields with many missing values. Add the **Filter** node by clicking on the **Field Operations** menu item in the Node palette and dragging the **Filter** node onto the canvas underneath the **Table** node. If the Node Palette is not visible, click on the Node Palette icon 
☐ first. Connect the **Titanic.csv** node to the **Filter** node. The canvas should appear as below.



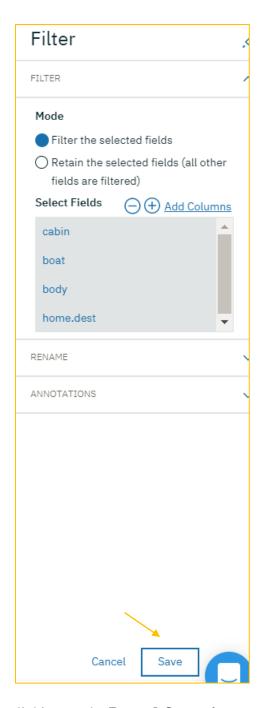
3. Double click on the **Filter** node. Click on the **Filter** dropdown. In the Filter panel, click on **Add Columns**.



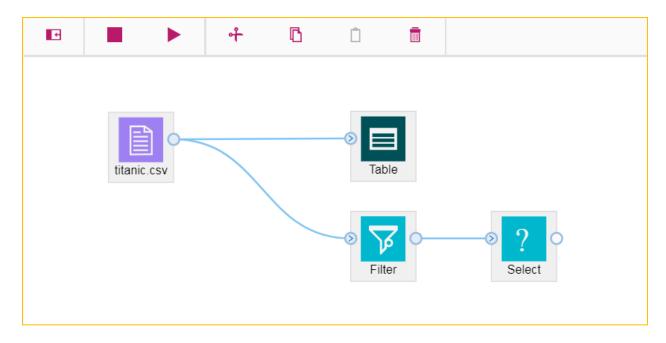
4. Click on the checkboxes adjacent to the **cabin**, **boat**, **body**, and **home.dest** fields, and then click on **OK**.



5. Click **Save** on the Filter panel.

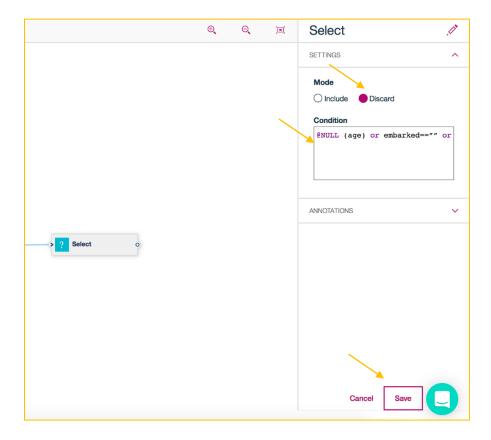


6. Add a **Select** node by clicking on the **Record Operations** menu item in the Node palette, and then dragging the **Select** node to the canvas to the right of the **Filter** node. Connect the **Filter** node to the **Select** node. If the Node Palette is not visible, click on the Node Palette icon irrst. The canvas should appear as below.

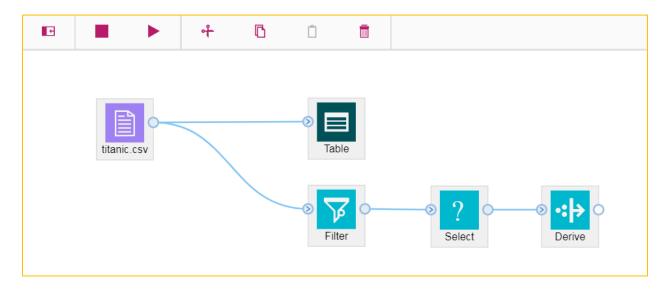


7. Double click on the **Select** node. Click on the **Settings** dropdown. In the **Select** panel, click on the **Discard** radio button, and re-type (or cut and paste) in the code shown below in the **Condition text box**, and then click **Save**.

@NULL (age) or embarked=="" or @NULL(fare)

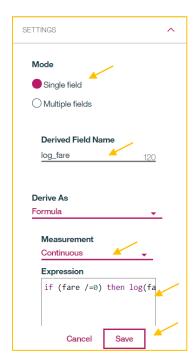


8. Add a **Derive** node to the canvas by clicking on the **Field Operations** menu item in the Node palette, and then dragging the **Derive node** onto the canvas to the right of the **Select** node. If the Node Palette is not visible, click on the Node Palette icon **E** first. Connect the **Select** node to the **Derive** node. The canvas should appear as below.



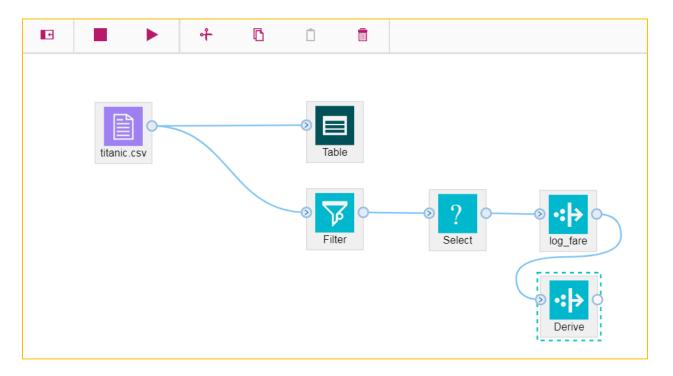
9. Double click on the **Derive** node. Click on the **Settings** Dropdown. Click on the **Single field** radio button, enter log\_fare for the **Derived Field Name**, select **Continuous** for the **Measurement**, enter the following code in the **Expression** text box, and click Save.

if (fare /=0) then log(fare) else 0 endif



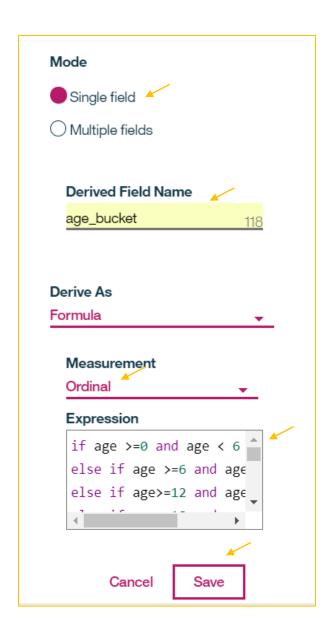
10. Binning of continuous fields is a technique sometimes used in preparing data for modeling. We will bin the age field, and the log\_fare field. Add a **Derive** node by clicking on the **Field Operations** menu item in the Node palette and dragging the **Derive** node on the canvas underneath the log\_fare **Derive** node.

If the Node Palette is not visible, click on the Node Palette icon first. Connect the log\_fare **Derive** node to the newly added **Derive** node. The canvas should appear as below.

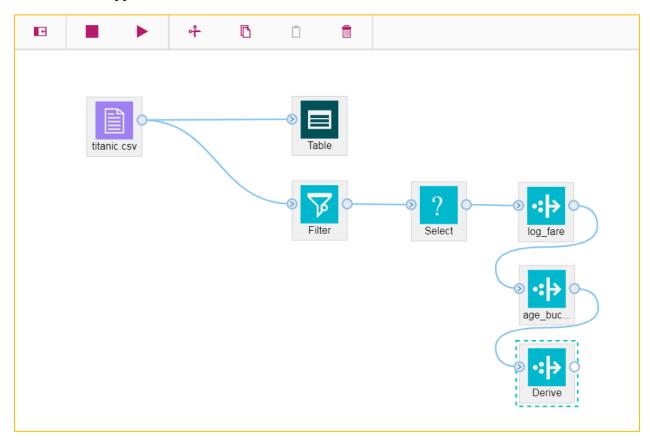


11. Double click on the **Derive** node. Click on the **Settings** dropdown. Click on the **Single field** radio button, enter age\_bucket for the **Derived Field Name**, select Ordinal for the **Measurement**, enter the following code in the **Expression** text box, and the click **Save**.

```
if age >=0 and age < 6 then 0
else if age >=6 and age < 12 then 1
else if age>=12 and age< 18 then 2
else if age>=18 and age <40 then 3
else if age>=40 and age <65 then 4
else if age>=65 and age<80 then 5
else 6
endif
endif
endif
endif
endif
endif
endif
```

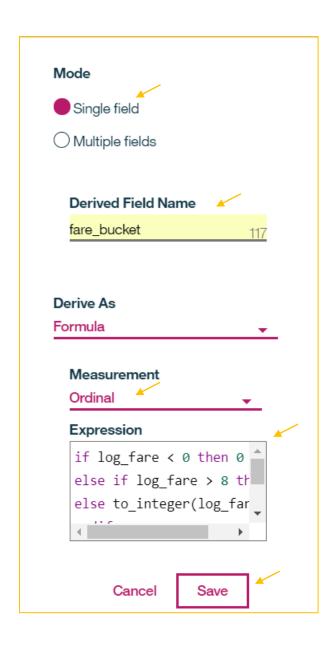


12. Add a **Derive** node by clicking on the Field Operations menu item in the Node palette and dragging the **Derive** node onto the canvas underneath the age\_bucket **Derive** node. Connect the age\_bucket **Derive** node to the newly created **Derive** Node. The canvas should appear as below.



13. Double click the **Derive** node. In the **Derive** panel, click on the **Single field** radio button, enter fare\_bucket in the **Derived Field Name**, click on Ordinal for the **Measurement**, enter the following code in the **Expression** text box, and click on **Save**.

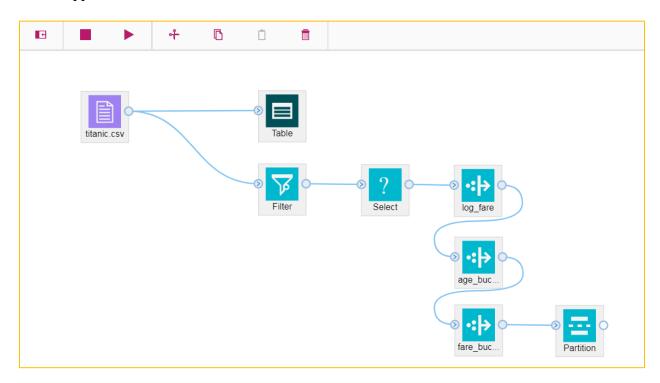
```
if log_fare < 0 then 0
else if log_fare > 8 then 9
else to_integer(log_fare)+1
endif
endif
```



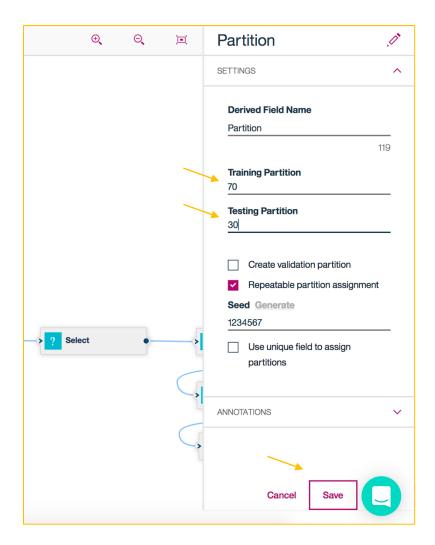
### Step 2.5 Modeling and Evaluation

Now that the data is prepared, we can start the modeling effort. First, we will add a **Partition** node to divide the data set into Training and Testing sets. In addition, a **Type** node is needed prior to modeling to type the new data fields that were created, and to assign roles to each field. Then we will add a **Logistic Regression** node, and use the Training set to train the model. Finally, we will add an **Analysis** node to evaluate the results.

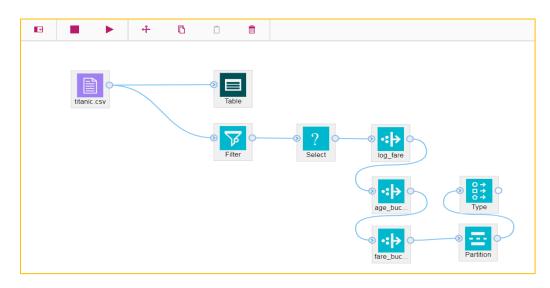
1. Add a **Partition** node by clicking on the Field Operations menu item in the Node palette and dragging the **Partition** node onto the canvas to the right of the fare\_bucket **Derive** node. Connect the fare\_bucket **Derive** node to the **Partition** node. The canvas should appear as below.



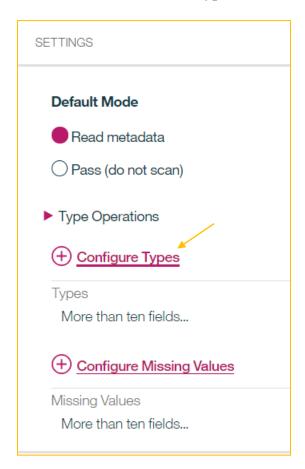
2. Double click on the Partition node. Set the **Training Partition** to 70 and the **Test Partition** to 30. Leave the other defaults and click on **Save**.



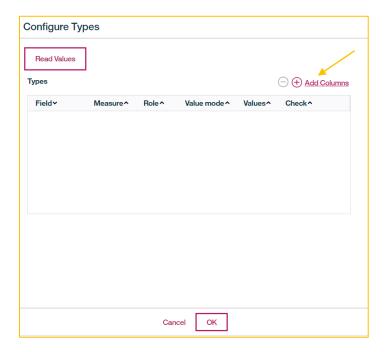
3. Add a **Type** node by clicking on the **Field Operations** in the Node palette and dragging the **Type** node onto the canvas above the **Partition** node. Connect the **Partition** node to the **Type** node. The canvas should appear as below.



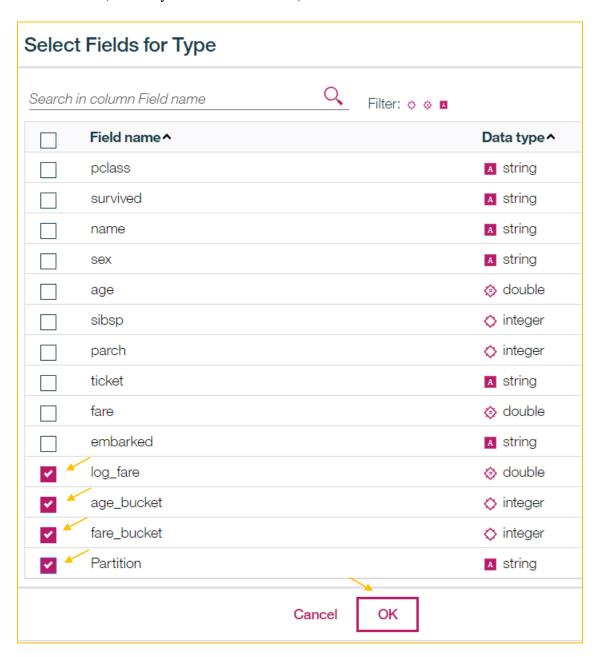
4. Double click on the **Type** node. Click on **Configure Types**.



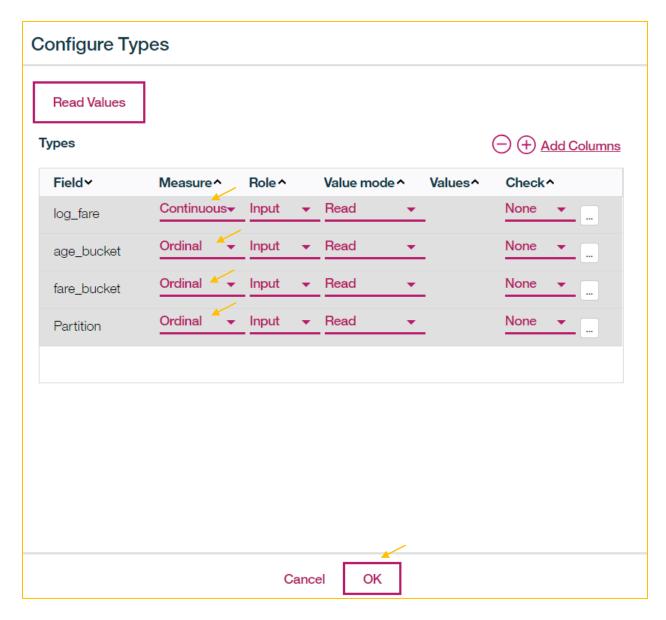
5. Click on Add Columns.



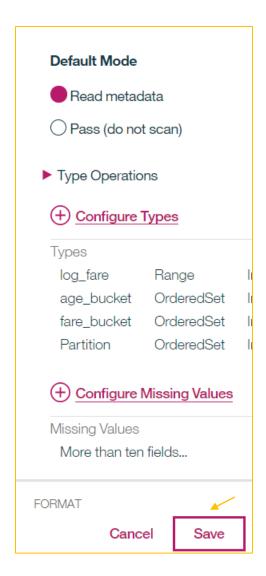
6. Click on checkboxes adjacent to the **log\_fare**, **age\_bucket**, **fare\_bucket**, and **Partition** fields (You may need to scroll down). Click on **OK**.



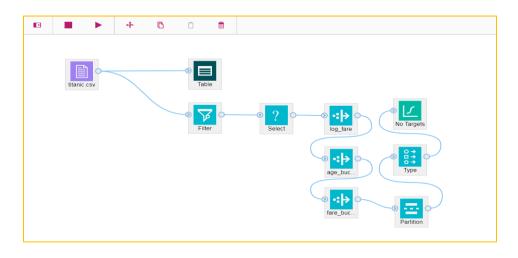
7. For the **Partition** field, select **Ordinal** for the **Measurement**. For the log\_fare, select **Continuous** for the **Measurement**. For the fare\_bucket field, select **Ordinal** for the **Measurement**, and for the age\_bucket, select **Ordinal** for the **Measurement**, and click **OK**.



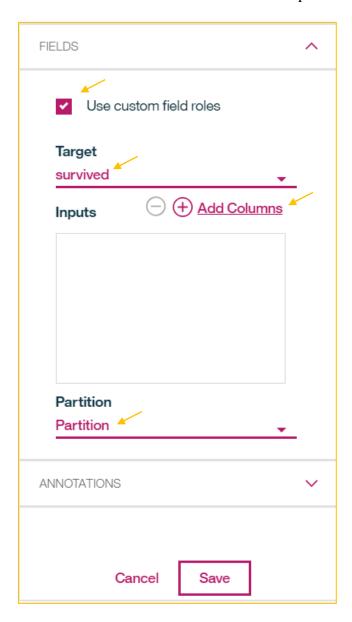
8. Click on **Save** 



9. Add a **Logistic Regression** node by clicking on the **Modeling** menu item in the Node palette and dragging the **Logistic** node onto the canvas above the **Type** node. Connect the **Type** node to the **Logistic Regression** node. The canvas should appear as below.



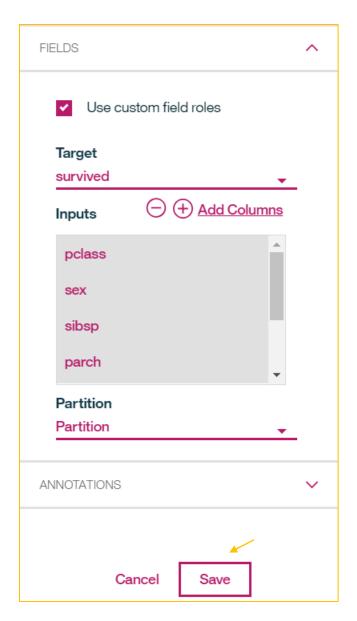
10. Double click on the **Logistic Regression** node. Click on the checkbox next to **Use custom field roles**, select **survived** for the **Target**, select **Partition** for the **Partition**, and click on **Add Columns** to add the input fields.



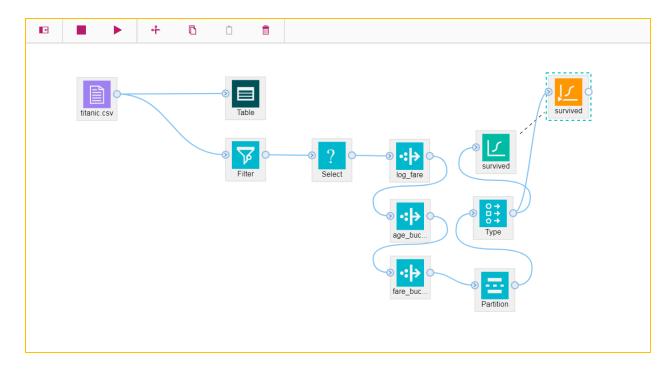
11. Click on the checkboxes next to pclass, sex, sibsp, parch, embarked, age\_bucket, fare\_bucket fields (you may have to scroll down), and then click OK.

Select	Fields for No Targets			
Search i	n column Field name	Q	Filter: ♦ ♦ ■	
	Field name ^			Data type ^
<b>~</b>	pclass			string
	name			string
<b>~</b>	sex			string
	age			double
<b>~</b>	sibsp			integer
<b>~</b>	parch			integer
	ticket			string
	fare			double
<b>~</b>	embarked			string
	log_fare			double
<b>~</b>	age_bucket			integer
<b>V</b>	fare_bucket			integer
		Cancel	OK	

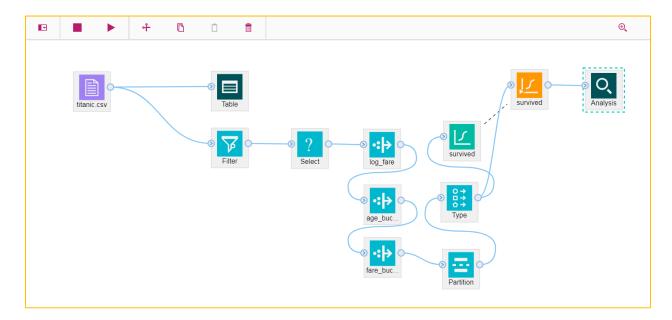
12. Click Save.



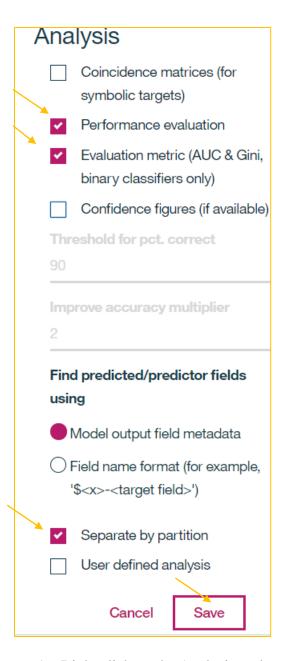
13. Right click on the **Logistic Regression** node and then click **Run**. A **Logistic Regression** "nugget will be created" connected by a dotted line to the **Logistic Regression** node. Drag the nugget and place it above the **Logistic Regression** node. The canvas should appear as below.



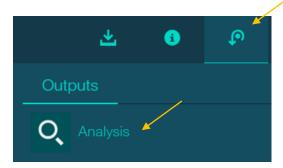
14. Add an **Analysis** node by clicking on the **Outputs** menu item in the Node palette and dragging the **Analysis** node onto the canvas above the nugget icon. Connect the nugget icon to the **Analysis** node. The canvas should appear as below.



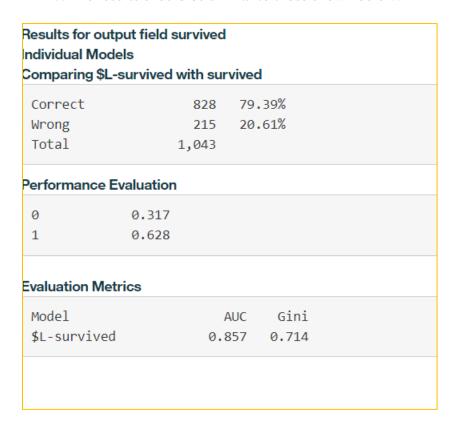
15. Double click on the Analysis node. Click on the **Settings** dropdown. Click on the **Performance evaluation** checkbox, **Evaluation metric** checkbox, uncheck **Separate by partition**, and click on **Save**.



16. Right click on the Analysis node and select Run. After completion, click on the Output icon and then double click on the Analysis link.



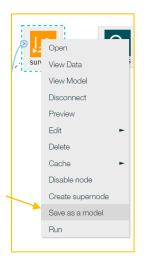
17. The results should be similar to those shown below.



# Step 2.6 Saving a Model

Now that we have created and evaluated a model, we will save the model as an asset. This saved model can be deployed at a future date, removing the need to recreate the same model from scratch.

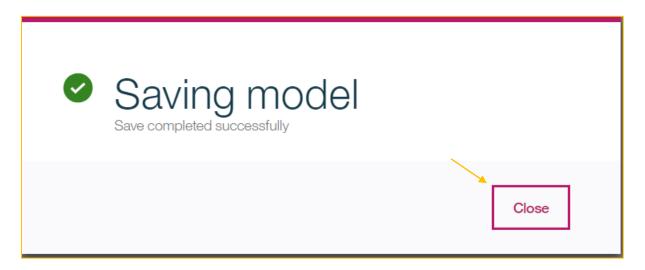
1. Right click on the Generated Model and then click on **Save as a model**.



2. Type in "Titanic-SPSS" as the Model Name and click Save.



3. Click Close.



4. Navigate to your project's **Models** page to see the saved Titanic-SPSS model. This model can then be deployed.

