



# Lab: Developing Analytics in a Visual Data Mining Workbench

*June 30th, 2018*

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Some features, like scripting, looping, and open source integration are not yet available in Modeler/DSX. Please check with the lab instructor if you would like to see the complete list of features that are not yet available.

*Note: Modeler Text Analytics will not be migrated to DSX. In DSX Text Analytics capabilities are provided by **Watson Explorer** (WEX).*

Modeler in DSX does not yet support all the same data sources as Modeler. See Compatibility Reports (**Reference** section) for the list of supported data sources.

Modeler in DSX uses JDBC drivers for supported data sources (included with DSX). SQL pushback is enabled by default for all supported database data sources. At this time there is no visual indicator for SQL pushback. Use Modeler documentation to understand which nodes can be pushed back.

Deployment is done in DSX **Deployment Manager**. We discuss flow deployment in more detail in the **DSX Deployment** hands-on lab.

## Introduction to Modeler for new users


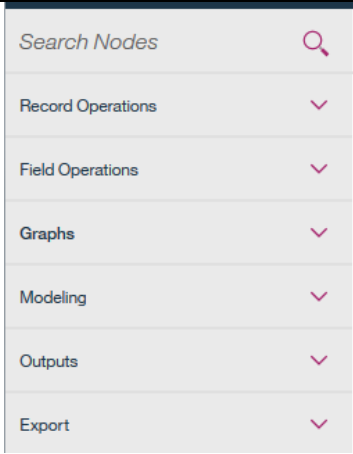





**SPSS Modeler** is a visual data mining workbench. Modeler can be used to complete all tasks of the analytic application development

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

Deployment is done in DSX **Deployment Manager**. We discuss flow deployment in more detail in the **DSX Deployment** 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.

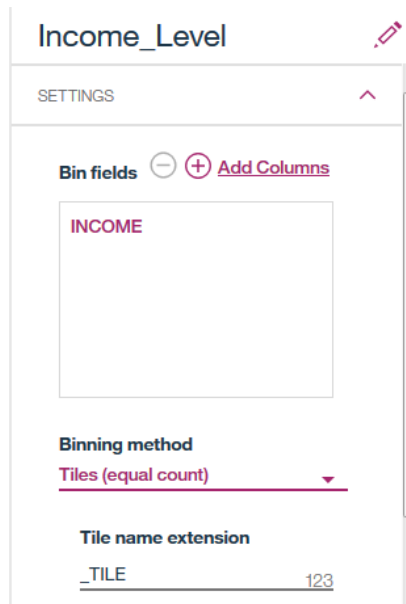
The data sources are purple		
Data preparation operations are blue		
Algorithms are green		
The models that are created based on algorithms are orange		
Different types of output (graphs, tables, external files) are black		
The nodes with a star icon are called “supernodes” because they contain several nodes. Supernodes are used for visual organization of the flow.		

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).

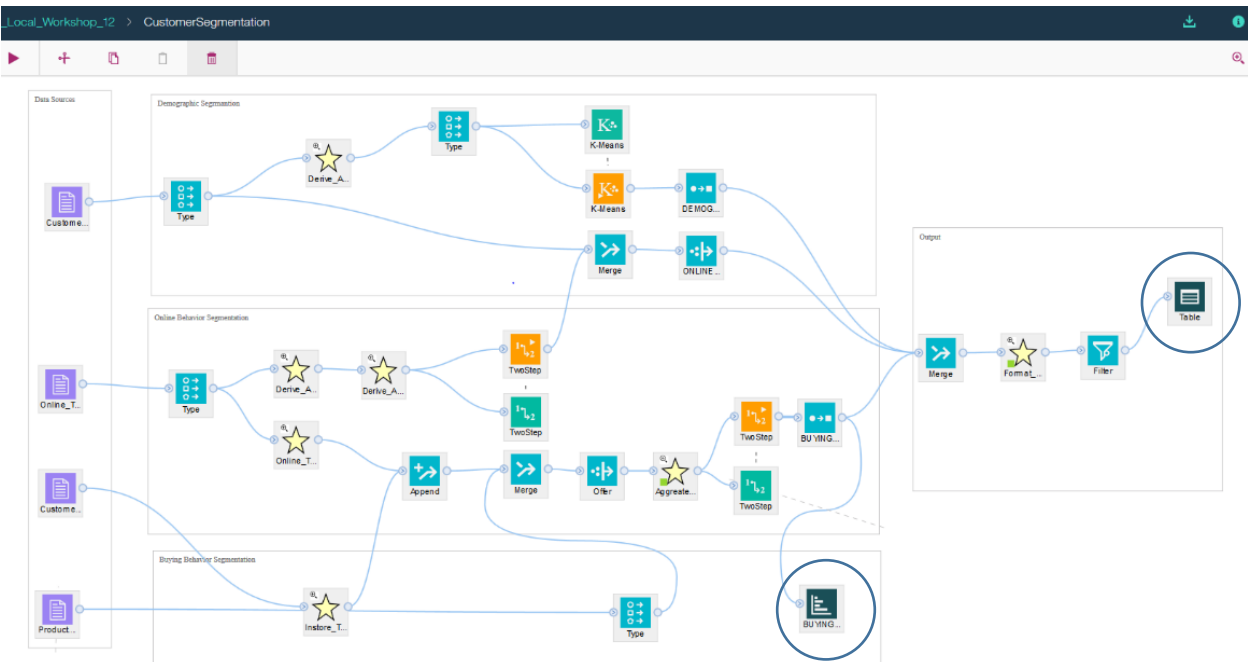
All visual nodes in Modeler have editable properties. You can bring up the properties view by double clicking on the node. Nodes should be edited after they have been connected to the previous node because Modeler is aware of data definitions, types, and manipulations

that happen in the flow. For example, if we use a **Derive** node to create a new field, the next node after it, a graph, can automatically detect this node.

### Settings for a Binning node



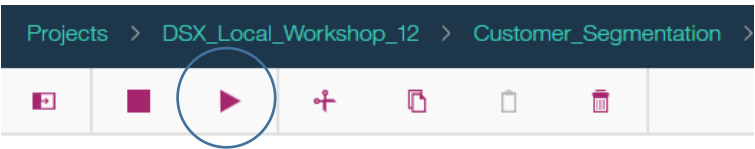
The node that doesn't have any outgoing connections is called a "terminal node". While any node can be a terminal node, typically a terminal node is a final step in a process: it may be an export data source, a visualization, an evaluation node, or a table that displays output. As shown in this example, the flow can have several terminal nodes.



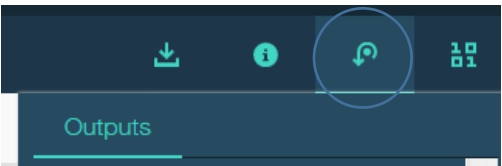
When working on the flow, we can execute different parts of the flow. If we click on the arrow icon in the menu bar, it will run all nodes of the flow. If the flow has output

(visualizations, tables), they will be available in the **Output** tab of the Modeler UI. If the flow builds a model, it will build the model node or replace the existing one.

*Run entire flow*

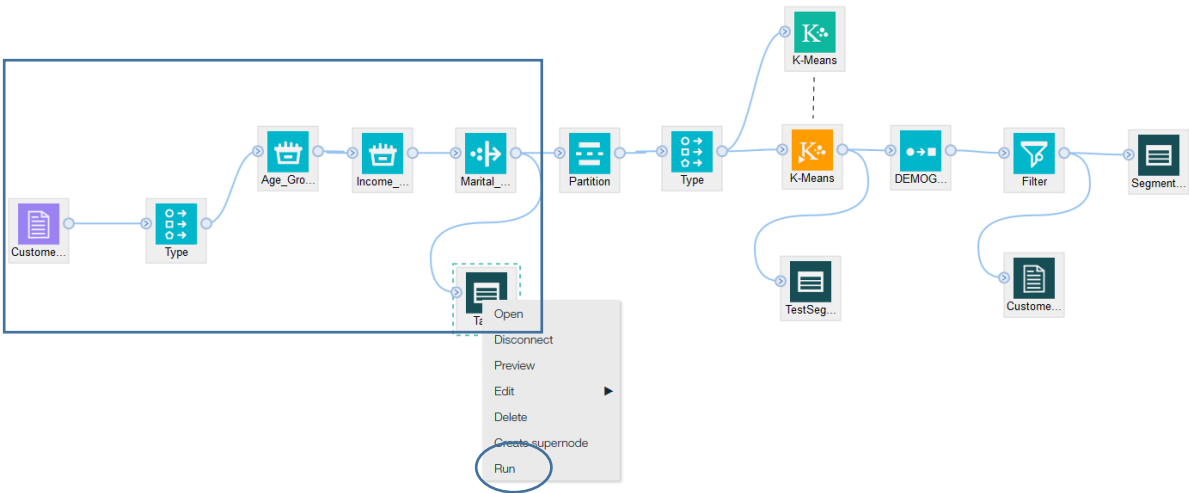


*Outputs tab*



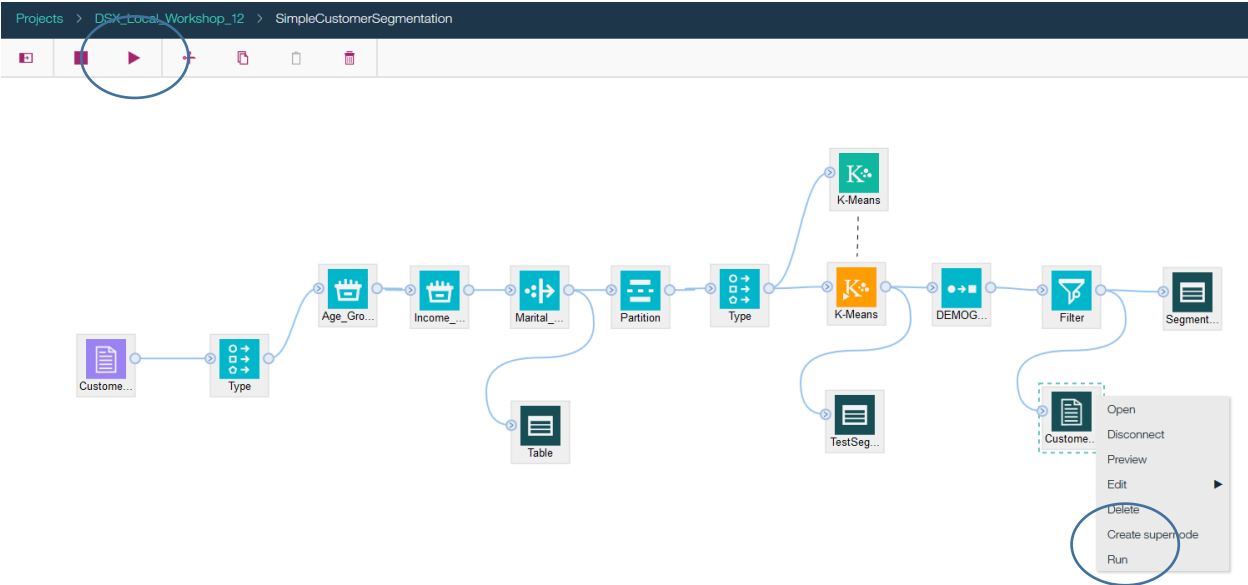
When a flow has several terminal nodes, it has several “branches”. We can run each branch by selecting the terminal node and running it.

*Selecting Run in the node menu runs the flow branch up to the selected node.*



If the flow has an **Export** node (a data source), when the entire flow or the export node branch is executed, data is written to the data source.

*Use the Run button on the menu or on the selected Export node to write data to an external data source.*



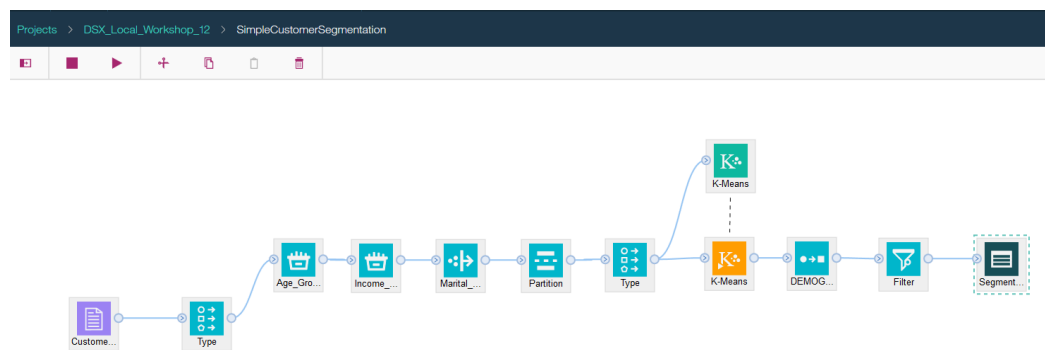


## Part 2: Review and edit a flow

In this section we will review a previously created flow, make changes, and run it.

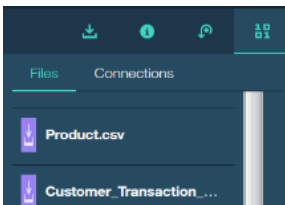
1. Log in to a **DSX Local cluster**.
2. If you haven't already imported *DSX\_Local\_Workshop\_12* project, import it.
3. Switch to the **Assets** tab and click on *SimpleCustomerSegmentation* flow to open it.

This flow performs several tasks to segment each customer into a demographic cluster. This type of segmentation can be included in a more sophisticated analytics workflow to create a customer profile. A complete customer profile can then be used to create more effective marketing campaigns.

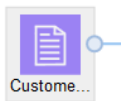


4. First, locate the data source in the flow. If we need to add new data sources, we can select them from the data sources view. Click on the **Data Sources** icon to view available data sources.

*Note: IBM publishes a compatibility report which shows the data sources supported by Modeler in DSX. See the **Reference** section.*



We have 1 data source in this flow, which contains demographic information about the customer.



- If you want to preview data in the data source, click on the data source and select **Preview**.

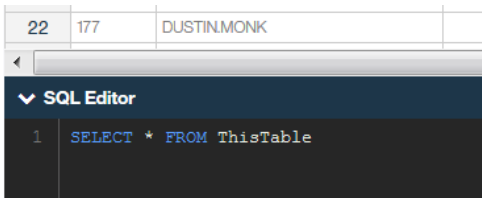
Preview typically loads the first 10 rows of data.

CUST_ID	REGISTRATION_ID	NAME	AGE	GENDER	MARITAL_STATUS	PROFESSION
292	MICHAEL.TOWB		33	F	M	Private Sect
305	TREVOR.KODJO		33	F	M	Professor
563	BILLY.POLO@EMAIL		33	F	M	Doctor
667	ADRIAN.DERN@EM		33	F	M	Engineer
1157	CHARLIE.HOUSE@E		33	F	M	Not employe
273	HELENA.KODJO		34	M	M	Professor

- Next, we'll take a look at some of the data understanding features. Right mouse click on the data source, *CustomerDataSegmentation*, and select **View Data**.

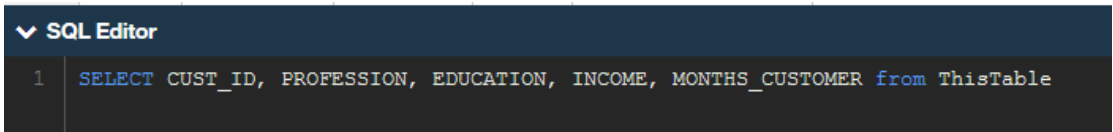
By default, the data is loaded in a spreadsheet format. Click on the column headings to sort data in the column.

If you would like to remove some columns from the view, click on the **SQL Editor** link at the bottom of the page.

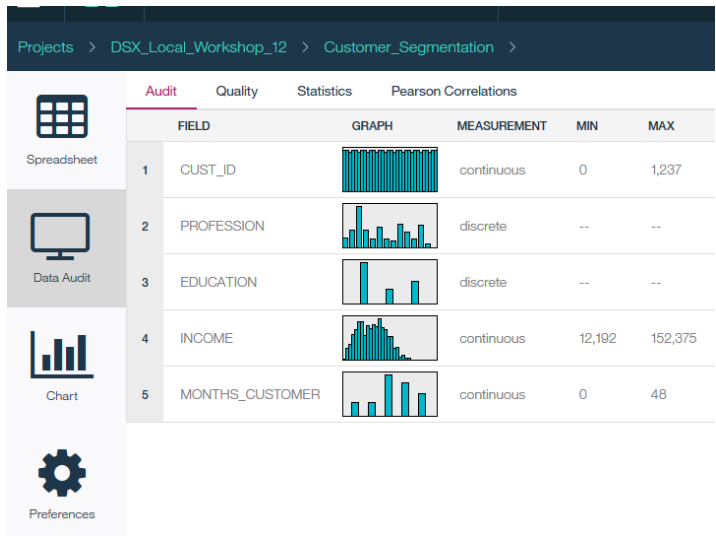


In this case "SQL" refers to the syntax that will be used. In our example we are working with data in a flat file. This editor is used to modify the view only, it does not persist the changes.

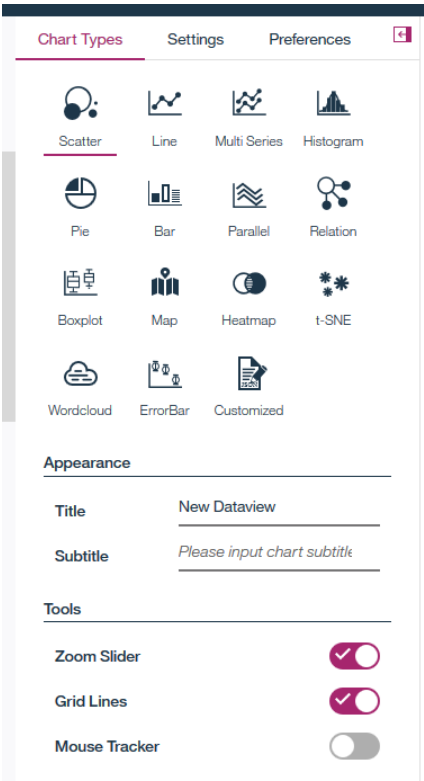
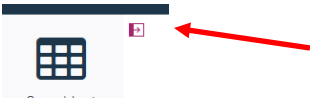
Try filtering out a few fieds, for example, select only the following fields. Click the arrow in the right corner to run the script.



Switch to the **Data Audit** view and review statistics for each column.

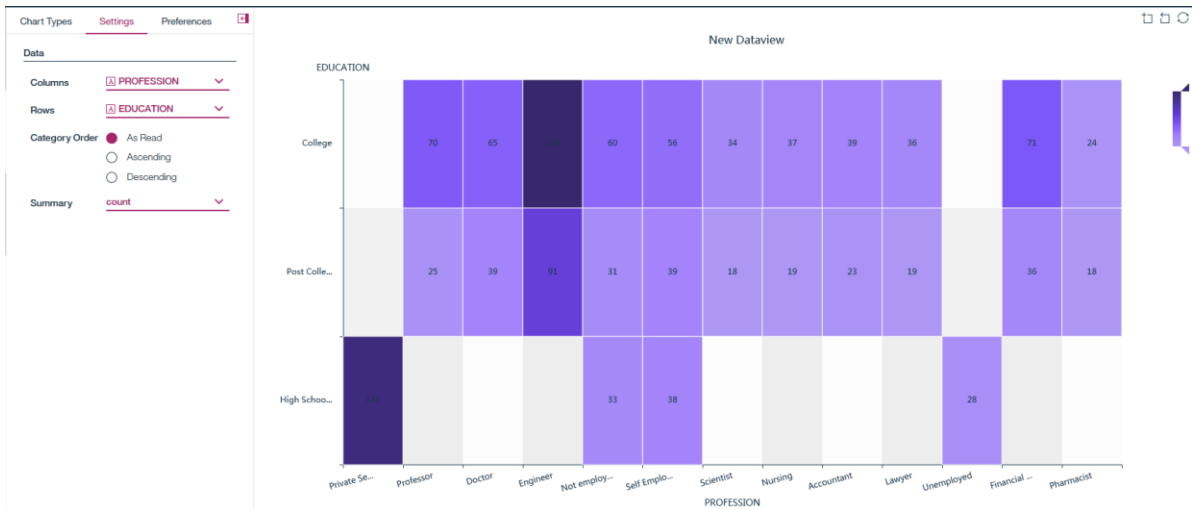
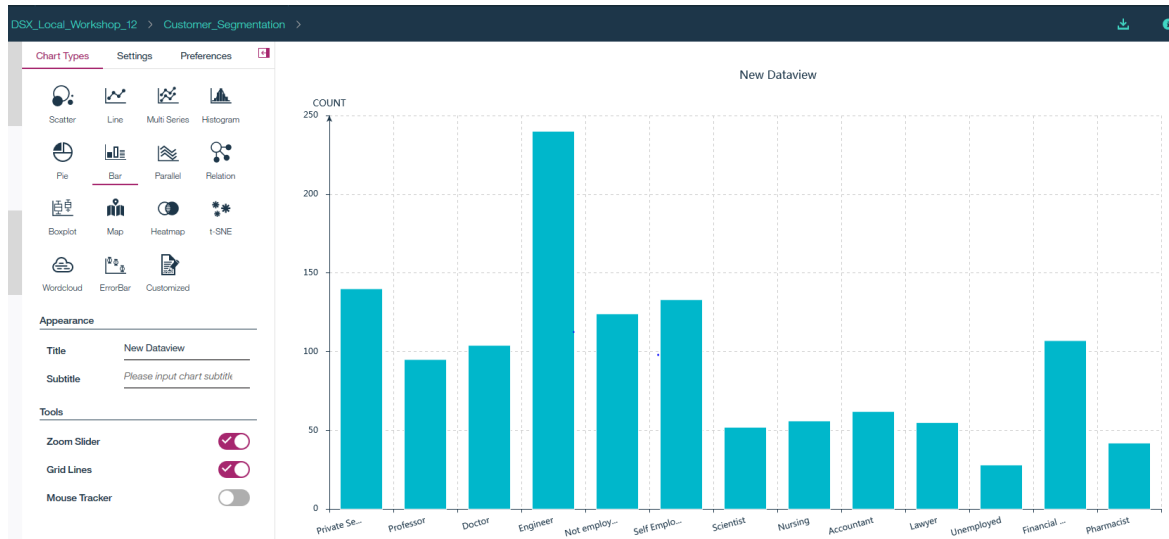


Switch to the **Chart** view and try creating a few charts. Bring up the chart wizard by clicking on the small arrow icon to the right of the Spreadsheet icon.



Try creating some charts, for example, create

- A bar chart of profession
- A heatmap of education and profession



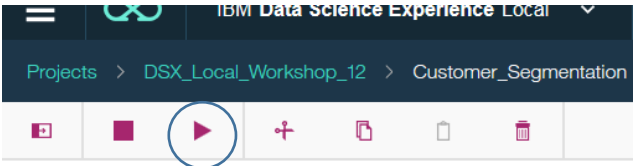
You may have noticed that not all fields are displayed for some graph types. Modeler automatically displays only the fields that are applicable for a specific graph type.

The top right corner of the graphs view provide the option to save the image.

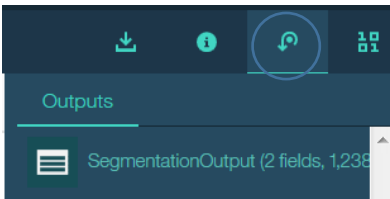


The majority of operations in the flow are data preparation. We create 1 model, a K-means segmentation model, and one output, a table, which shows the results of segmentation.

- Let's run the stream, and then we'll take a closer look at the implementation. Click the arrow icon to run the entire stream.



- The output is accessible through the **Output** view. The output table displays the segmentation results.



- Double click on output to view it. The output shows customer id and demographic cluster (the value that was generated in the flow).

CUST_ID	DEMOGRAPHIC_CLUSTER
292	MIDDLE INCOME
305	MARRIED WOMEN
563	HIGH INCOME FEMALE
667	MARRIED WOMEN

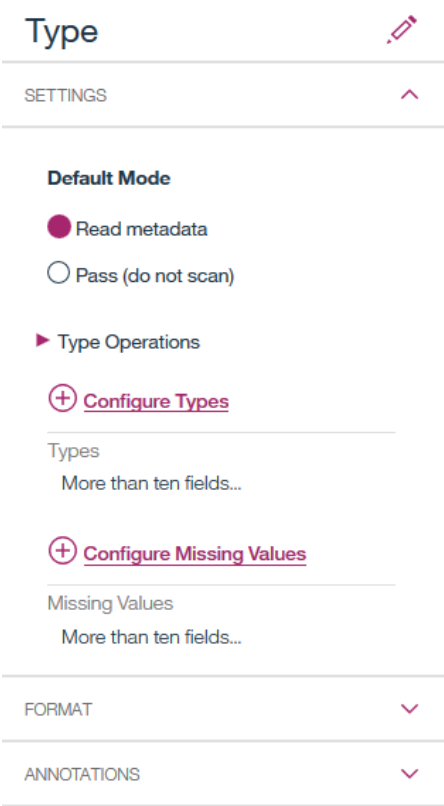
A "Table" is used to display data in Modeler. Later we will add an **Export** node which will write data into an external data source.

- Next, let's review some of the frequently used nodes.

The data source node is connected to a **Type** node. Since Modeler node functions can be different depending on a data type (like we've seen in the example of graphs), it's important to identify the data type before we apply any operations to data.

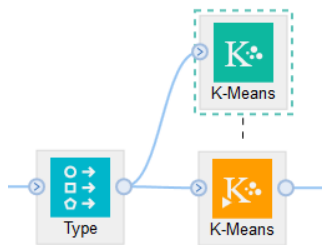
Modeler performs "auto typing", and in most cases the default node settings don't need to be modified. However, if users want to modify this node, it can be done through the properties view.

Double click on the first **Type** node connected to the *CustomerDataSegmenation.csv* data source and review the settings. Notice that you can do some basic data transformation in this node (Click **Configure Types**).



We also have a second **Type** node in the same branch, right before the K-means algorithm node. When the **Type** node is placed right before the algorithm, it's used to specify fields that will be used for modeling, the target fields, and fields that should be filtered out.

Double click on the second **Type** node, and take a look at the roles that have been specified for modeling (Click on **Configure Types**).



## Configure Types

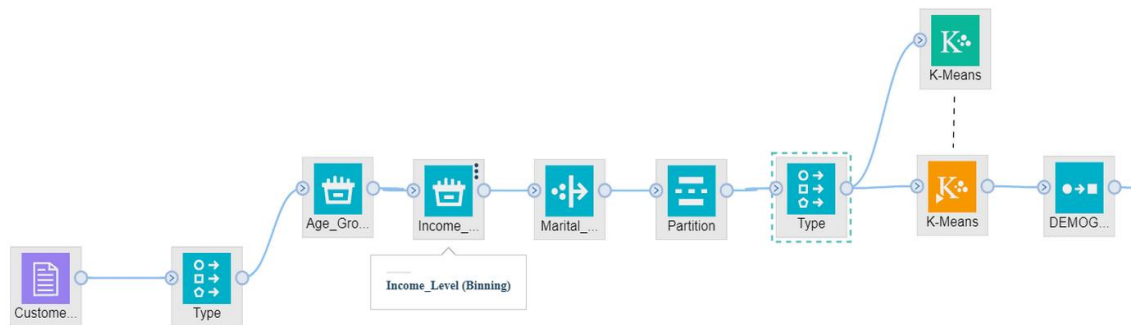
Read Values

Types ⊖ ⊕ Add Columns

Field^	Measure^	Role^	Value mode^	Values^	Check^
CUST_ID	Continuous	Record	Specify	0, 1237	None
Marital_Status_Derived	Flag	Input	Specify	M, S	None
MARITAL_STATUS	Nominal	None	Specify	M, S, U	None
PREDICTOR_C_ON-LINE_RESPONSE	Continuous	None	Specify	0, 1	None
PREDICTOR_D...	Continuous	None	Specify	0, 000, 0000	None

After the first **Type** node we have several data preparation operations that are common in the analytics process, such as binning, deriving new variables, and partitioning data before modeling.

If you position the mouse over the node, it will show a tooltip with the node type.



Similar to the **Type** node, you can view the settings for each node by double clicking on it or selecting **Open** from the right click menu.

If you view the **Settings** for the *MaritalStatus\_Derived* field, you'll see that it contains a formula. The formula derives the marital status if it's empty.

**Measurement**  
Categorical

**Expression**

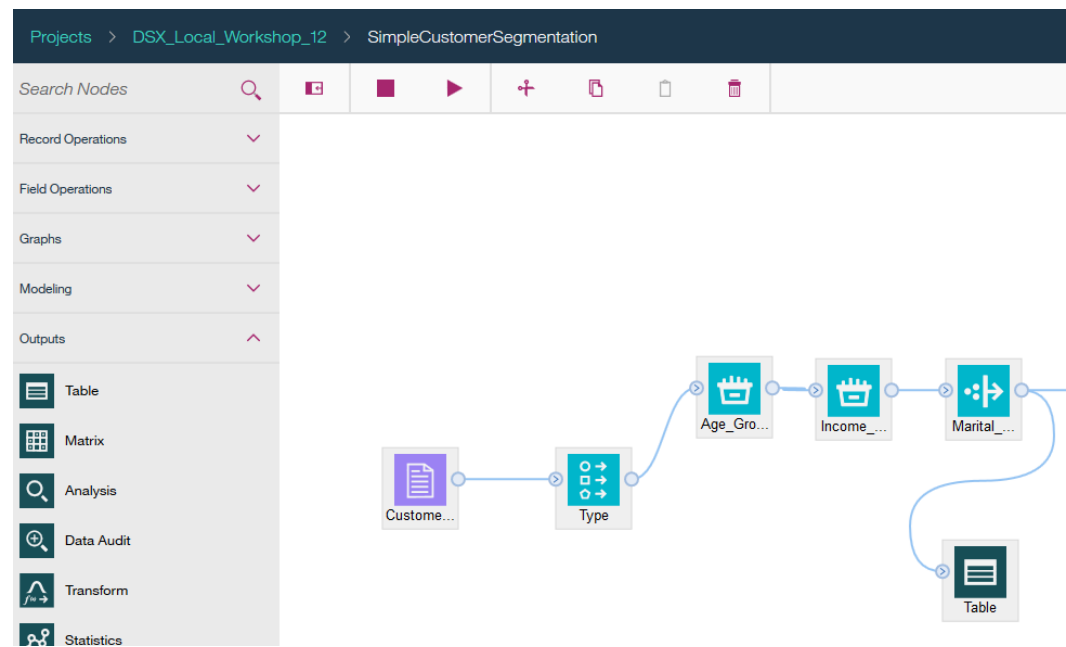
```
if (MARITAL_STATUS = "S"
```

This formula is called "Expressions" in Modeler. In the future releases, Modeler in DSX will have an "Expression Builder", which supports building expressions by selecting fields and actions in a dialog box (i.e. you will not have to write the code). This capability is already available in Modeler desktop.



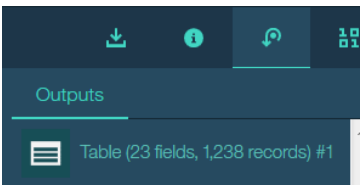
At this time, you can use documentation to look up syntax for expressions:  
[https://www.ibm.com/support/knowledgecenter/en/SS3RA7\\_18.1.1/modeler\\_mainhelp\\_client\\_ddita/clementine/clem\\_function\\_ref.html](https://www.ibm.com/support/knowledgecenter/en/SS3RA7_18.1.1/modeler_mainhelp_client_ddita/clementine/clem_function_ref.html)

13. Attach a **Table** node (from **Output** tab) to the **Marital\_Status\_Derived** node.



14. Right mouse click on the **Table** node and click **Run**.

When we select **Run** from a node menu, all nodes from the source until that node are executed. Output is accessed through the same **Output** view.

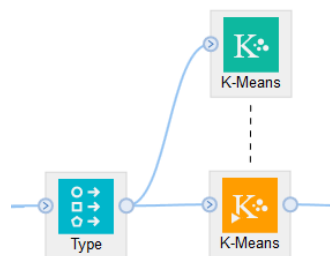


Double click on the table, and scroll all the way to the right. The fields that were derived by the data preparation nodes are appended at the end of the dataset. *AGE\_TILE5* and *INCOME\_TILE5* were created by the **Binning** nodes, and *MARTITAL\_STATUS\_DERIVED* was created by the **Derive** node.

AGE_TILE5	INCOME_TILEN	MARITAL_STATUS_DE
3	1	M
3	2	M
3	3	M
3	1	M

Select **SimpleCustomerSegmentation** to return to the model view.

- Next, let's review the algorithm and the model nodes. The green icon is the algorithm which was selected from the **Models** palette, and the orange icon is the model that was created based on provided dataset.



Double click on the green **K-means** icon and review the properties. The **Fields** are empty because the algorithm is using field "roles" (input, target, none) from the **Type** node in front of it. This is typical implementation in flows – to specify field roles in a separate **Type** node.

K-Means

FIELDS

☐ Use custom field roles

Inputs

–

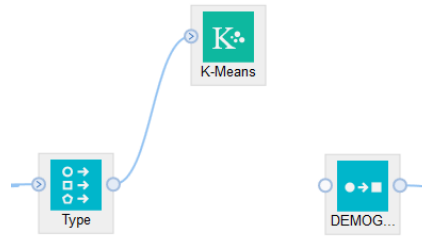
+

Add Columns

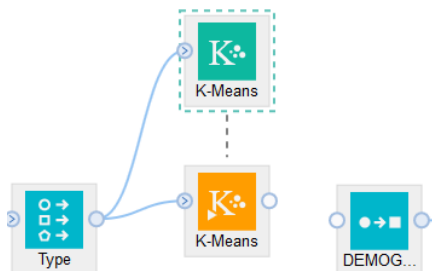
If you wish, review the **Type** node to understand which fields are used for training this model (*Hint: double click on **Type** node, then select **Configure Types***).

The rest of the settings in the algorithm were provided by default. An experienced data scientist can use **Build Options** and **Expert Options** for each model type, but in most cases the default values provided by Modeler don't need to be changed.

Let's delete the generated model and build a new one. Select the orange **K-means** icon and delete it.



Right mouse click on **K-means algorithm** (green icon), and select **Run**. This generates the K-means model.



16. Right-click on the generated model and select **View Model** to review model details. This information helps data scientists to understand the model.

Projects > DSX\_Local\_Workshop\_12 > SimpleCustomerSegmentation > K-Means

K-Means Clustering Model ⓘ

Model Information

Predictor Importance

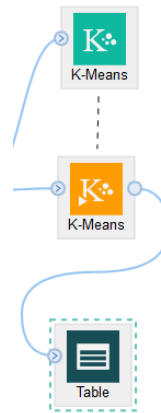
Cluster Sizes

Model Information ⓘ

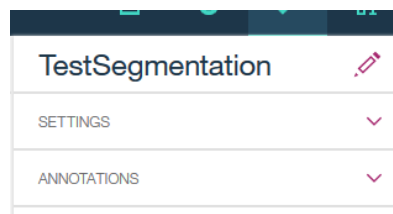
Algorithm	K-Means
Model Class	Center Based
Number of Features	7
Distance Measure	Euclidean
Number of Clusters	6

Return to the model view.

17. Add a table to the **K-means** model node. (Hint: open the Palette, then open Outputs and drag a table icon onto the canvas. Then link the table to the model)



Double click on the table and name it *TestSegmentation*. If we don't rename the table, it may be confusing to find it in the **Output** view. Make sure to save the changes for the Table node.



Run the flow up to the table by selecting **Run** from the **Table** right click menu. View the output.

The model generated the cluster field, which was appended to the end of the data source.

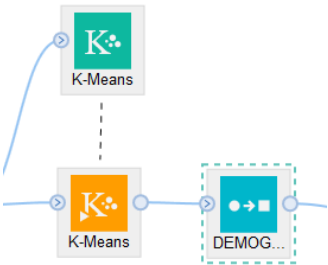
PARTITION	SKM-K-MEANS
1_Training	cluster-1
1_Training	cluster-5
1_Training	cluster-3
2_Testing	cluster-5

The cluster information without explanation may not be useful to a marketing professional, that's why we need to add a Reclassify node, to give the clusters an easily understandable name.

*Note: Modeler desktop segmentation algorithms provide an easy way to understand the composition of the cluster, and this capability is currently being ported to DSXL. We implemented a separate Modeler stream, ClusterUnderstanding, to explain how it can be*

done at this time. Please check with your instructor if you would like to learn more about the approach we used.

18. Connect the generated model to the **Reclassify** node next to it. The **Reclassify** node assigns a more meaningful name to a cluster.



Double click on the **Reclassify** node to review how clusters are renamed.

Reclassify Field

SKM-K-Means

New Field Name

DEMOGRAPHIC\_CLUSTER

Get values

Copy

Clear new

Automatically Reclassify

Values

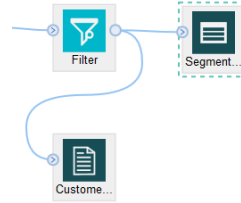
ORIGINAL VALUE^

NEW VALUE^

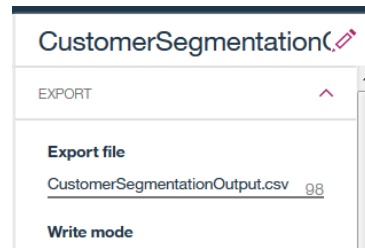
cluster-1	MIDDLE INCOME SCHOOL EDUCATED
cluster-2	HIGH INCOME SINGLE
cluster-3	HIGH INCOME FAMILY
cluster-4	MIDDLE INCOME SINGLE
cluster-5	MARRIED WOMEN

19. The next node is the **Filter** node. This node simply removes the fields that we don't need in the final output. Double click on the **Filter** node and review its properties.
20. At this time the final output of the flow is displayed in a **Table** node, which is internal to Modeler. Let's add an **Export** node, which will write output to a file or a database datasource.

Connect a **Flat File** node (from the Export palette) to the **Filter** node.

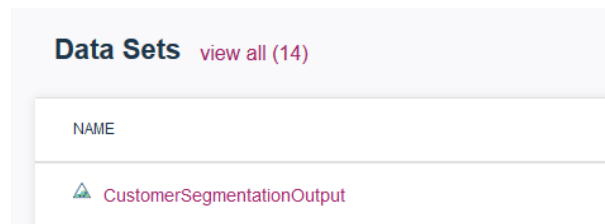


Edit the **Flat File** node to give it a name (change in properties) – for example, *CustomerSegmentationOutput.csv*. It's important to add .csv extension to enable preview and download capabilities in DSX projects.



21. Run the entire stream by selecting the **Run** icon from the main menu bar. Navigate to the DSX project view.

The flow generated the output file which you can preview and export (use the vertical ellipses button).



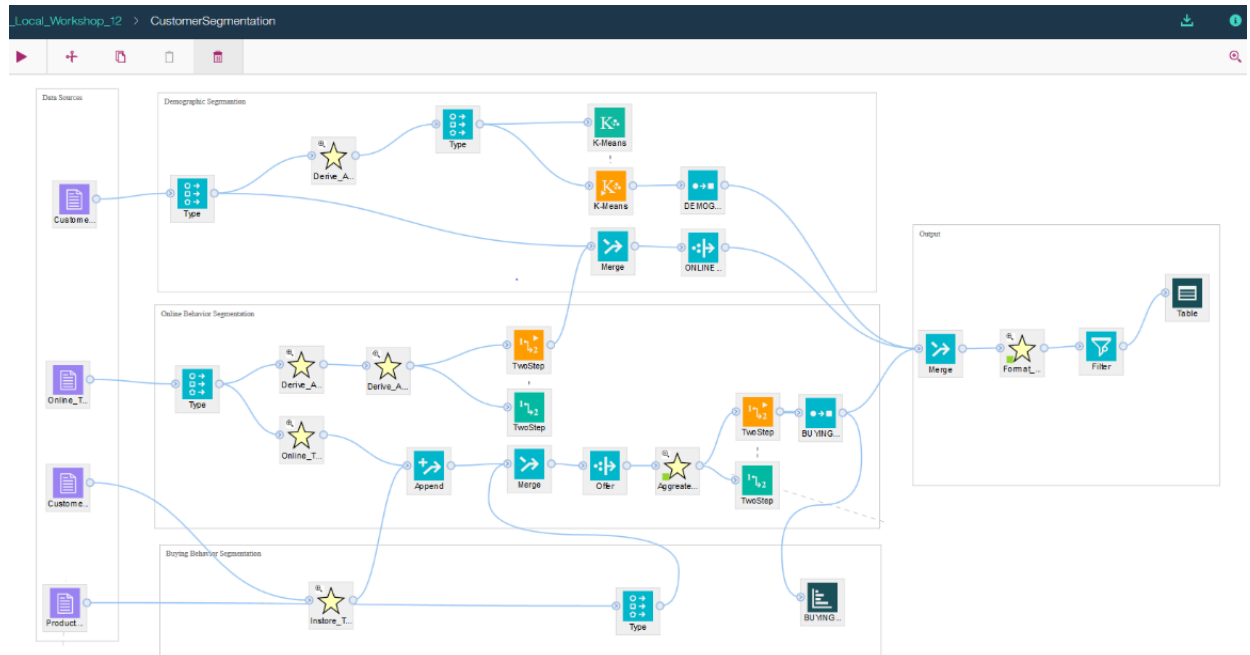
If we add a "Data Connection" type of Export node in Modeler, then the flow will write to the specified database source.

**We have finished reviewing a simple Modeler flow. Now let's review a more complicated one.**

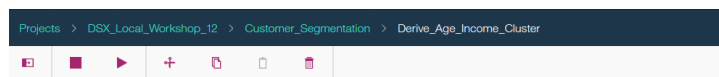
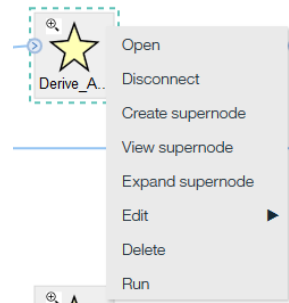
22. Switch to the **Assets** tab and click on *CustomerSegmentation* flow to open it.

This flow performs several tasks to derive the following information for each customer: Customer Lifetime Value (LTV), buyer category, demographic cluster, online behavior cluster, and buying behavior cluster. This information can help companies run more effective marketing campaigns.

The top part of the flow performs demographic segmentation, which we already reviewed.



The yellow icons with a star are called “*supernodes*”. They are used for visual organization in a flow. Supernodes contain several other nodes. Click on one of the supernodes and select **View Supernode**. You can add additional nodes inside of a supernode.



To create a supernode, you can select multiple nodes with the **Shift** key, and use the right click menu to create the supernode. You can also “expand” the supernode using the right click menu option inside the supernode view.

Take time to review this flow and let the instructor know if you have questions about nodes or the analytics implementation.

Try running this flow and review the outputs such as SegmentationOutput. In addition to the demographic cluster, which we generated in the demographic segmentation flow, the flow creates additional fields that create a more complete customer profile.

CUST_ID	REGISTRATION_ID	LTV	DEMOGRAPHIC_CLUSTER	ONLINE_BEHAVIOR_CLUSTER	BUYING_BEHAVIOR_CLUSTER
		LOW VALUE	MIDDLE INCOME SINGLE	PRODUCT VIEWER	INSTORE MEDIUM
0.000		LOW VALUE	MIDDLE INCOME SINGLE	VISITOR	INSTORE MEDIUM
1.000	50524	HIGH VALUE	MIDDLE INCOME SINGLE	VISITOR	INSTORE LOW VALUE
2.000	226666	LOW VALUE	MARRIED WOMEN	VISITOR	INSTORE MEDIUM
3.000	254376	MEDIUM VALUE	MIDDLE INCOME SINGLE	VISITOR	ONLINE MEDIUM
4.000	288205	MEDIUM VALUE	HIGH INCOME FAMILI	VISITOR	INSTORE LOW VA
5.000	660891	LOW VALUE	MARRIED MEN	VISITOR	INSTORE MEDIUM

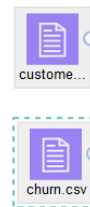


## Part 3: Create a new flow

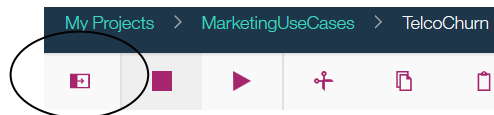
In this section you will create a Modeler flow in DSX/Modeler UI. The process is the same as creating a Modeler stream in **SPSS Modeler** desktop, the only differences are in the UI look and feel.

If you have worked on other hands-on labs in the DSX workshop – this Modeler flow implements the same use case as Telco Churn Jupyter and Zeppelin notebooks (available in the *DSX\_Local\_Workshop\_12* project).

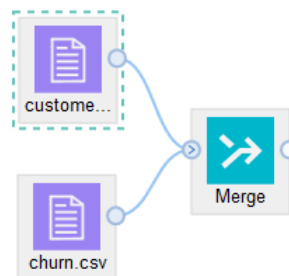
1. On the **Assets** tab scroll down to **SPSS Modeler flows** and click **add flow**.
2. Provide a unique stream name (for example, add your initials) *TelcoChurn\_el*. Click **Create**.
3. Drag and drop *customer* and *churn* data source to the canvas.



4. Right mouse click on the data sources in the canvas and select **Preview** to view the data. *Customer* dataset contains demographic and service usage information and *churn* dataset contains historical data showing if the customers has churned.
5. Click on the **Palette** icon and expand **Record Operations**.



6. Add the **Merge** node then connect the *customer* and the *churn* data sources to it.



7. Double click on the **Merge** node. Select *Keys* as the **Merge method**.

Merge

INPUTS

MERGE

Merge method

Keys

Keys

+

Add Columns

8. Click **Add Columns** and select the *ID* field. Click **OK** to return to the previous screen.

Select Fields for Merge

Search in column Field name

Filter:

Reset

<input type="checkbox"/>	Field name ^	Data type ^
<input checked="" type="checkbox"/>	ID	integer
<input type="checkbox"/>	Gender	string

9. Now the Merge screen looks like the following screenshot. Click **Save**.

Merge method

Keys

Keys

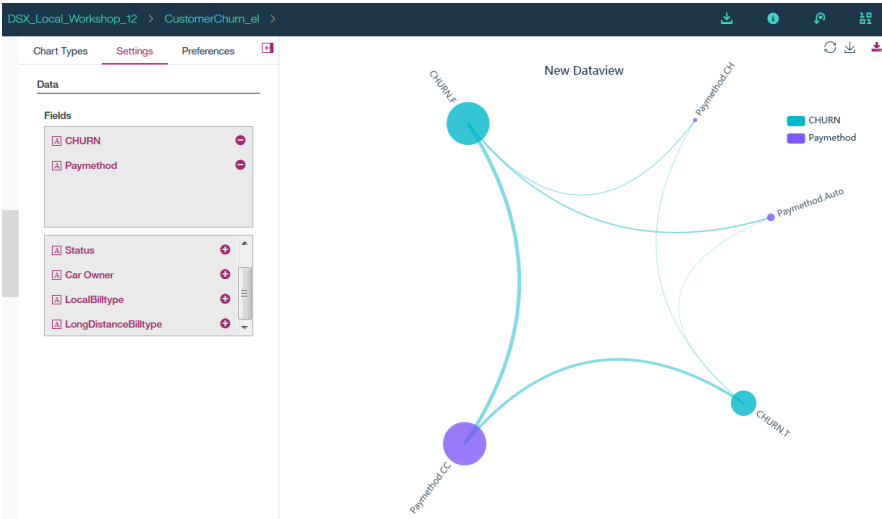
+

Add Columns

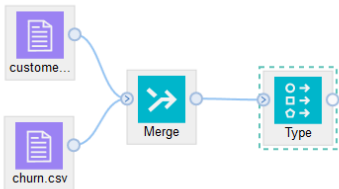
ID

Next, we can perform several data understanding tasks by selecting **View Data** from the right click menu of the **Merge** node.

Try several different steps, as we've done in the previous section. For example, you can create a relationship graph for churn and payment method. This graph shows that a small percentage of customers who use autopay have churned.



10. Next, we are going to build a model for predicting churn. Add a **Type** node from the **Field operations** and connect it to the **Merge** node.



11. Double click on the **Type** node and click **Configure Types**. In the **Configure Types** screen click **Read Values**.

Configure Types

Read Values

Types ⊖ ⊕ Add Columns

Field^	Measure^	Role^	Value mode^	Values^	Check^
Dropped	Continuous	Input	Specify	0.0, 4.0	None
Status	Nominal	Input	Specify	D, M, S	None
Usage	Continuous	Input	Specify	0.68, 361.88	None
Est Income	Continuous	Input	Specify	96.33, 120...	None
Paymentmethod	Nominal	Input	Specify	Auto, CC, CH	None

- Sort fields by name (so that it's easier to find them). Change the **Role** of *churn* to **target** and *ID* to **Record ID**.

This means that all fields will be used as input to modeling, with the exception of the *ID* field. *Churn* is the field we want to predict so set its role to **Target**.

Read Values

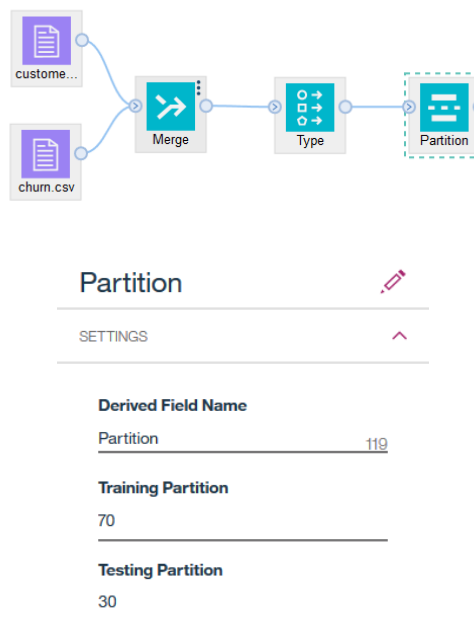
Types

⊖ ⊕ Add Columns

Field^	Measure^	Role^	Value mode^	Values^	Check^
Age	Continuous	Input	Specify	12.326667,...	None
CHURN	Flag	Target	Specify	F, T	None
Car Owner	Flag	Input	Specify	N, Y	None

Click **OK**, then **Save** to save the settings in the **Type** field.

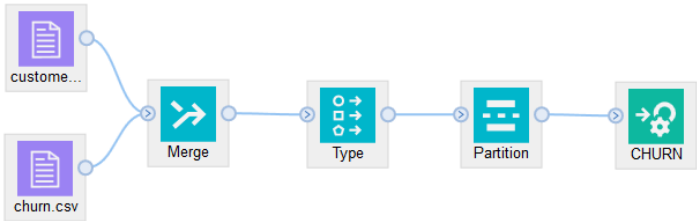
- From the **Field Operations** add the **Partition** node and set the **Training Partition** to 70 and the **Testing Partition** to 30. Save the changes in the node.



- Add the **Autoclassifier** node from the **Modeling** tab. **Autoclassifier** will automatically create several classification models and show model accuracy for the best models.

Modeler documentation explains which classifier models are built, and why some models are discarded (see **Reference**).

The **Autoclassifier** node is named "Churn" because we specified churn as the target field in the Type node.



- Right click on the **Autoclassifier** node (*CHURN*) and select **Run**.
- Model building will take a few minutes. When model building is done, you'll see a model node (orange icon on the canvas).

Right-click on the generated model and select **View Model**. The top three models generated are shown in the screenshot.

Projects > DSX\_Local\_Workshop\_12 > CustomerChurn\_el > CHURN

Auto Classifier ⓘ

Models

Auto Classifier - Models ⓘ

TARGET : CHURN

USE	ESTIMATOR	CLASSIFICATION ACCURACY	BUILD TIME (MINS)	NO. FIELDS USED	ACTIONS
	XGBoost Tree 1	96.142	< 1	15	
	Random Trees	96.142	< 1	15	
	MLP Neural Network	91.049	< 1	15	

- Click on ellipses next to *Random Trees* model and select **View**.

	Random Trees	96.142	< 1	<div> View Delete </div>
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Review model information. For example, the **Top Decision Rules** view is useful for understanding the factors that lead to customer churn.

Projects > DSX\_Local\_Workshop\_12 > CustomerChurn\_el > CHURN

Auto Model - CHURN

Random Trees ⓘ

Model Evaluation

Model Information

Records Summary

Predictor Importance

Top Decision Rules

Confusion Matrix

Top Decision Rules ⓘ

TARGET : CHURN

Table Contents: Top Decision Rules

Decision Rule	Most Frequent Category	Rule Accuracy	Ensemble Accuracy	Interestingness Index
<ul style="list-style-type: none"> <li>LongDistance &gt; 16.04</li> <li>Est Income &gt; 55860.0</li> <li>Children &gt; 1.0</li> <li>Status =</li> </ul>	F	1.000	1.000	1.000

By default ensemble scoring will be used. Some ensemble scoring settings can be changed in model properties. As you’ve seen in the model details view, models can be deleted (removed from the ensemble).

In some cases data scientists use the “auto” modeling nodes to determine the best model, and then they use that individual model from the models palette.

18. Attach **Table** and **Analysis** nodes to the model, and run each one by selecting **Run** from the right click menu.

*Note: if you select Run from the main menu, Modeler will run the entire stream, which will rebuild the model.*

Lab: SPSS Modeler in DSX

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Review the output. The table shows the predictions that were generated by the model (scroll to the right).

PARTITION	\$XF-CHURN	\$XFC-CHURN
1_Training	T	0.984
1_Training	F	0.819
1_Training	F	0.782
2_Testing	F	0.788
1_Training	F	0.974
1_Training	F	0.983

Model analysis output

Projects > DSX\_Local\_Workshop\_12 > CustomerChurn\_el > Analysis of [CHURN]

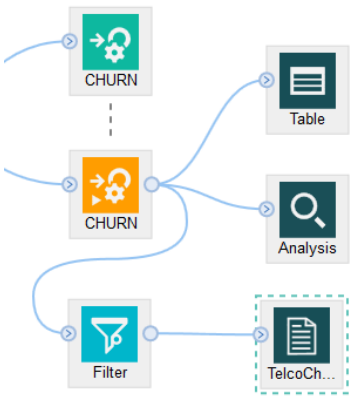
Results for output field CHURN

Comparing \$XF-CHURN with CHURN

'Partition'	1_Training		2_Testing	
Correct	1,393	98.24%	625	96.45%
Wrong	25	1.76%	23	3.55%
Total	1,418		648	

19. Finally, add a **Filter** and **Export (Flat File)** nodes.

- In the **Filter** node remove all fields with the exception of ID, predicted churn value (\$XF-CHURN), and confidence in the prediction (\$XFC-CHURN),
- In the **Export** node specify a .csv file as an export.



20. Run the file export branch: right mouse click on the flat file output node and select **Run**.

21. Navigate to the Project view and review the generated file.

## Preview - TelcoChurnScoring.csv

ID	\$XF-CHURN	\$XFC-CHUR
1	T	0.984006
6	F	0.818762
8	F	0.782236

You have finished building a Modeler flow in DSX.

You have finished developing a model to predict customer churn.



## Reference

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**Modeler in DSX documentation:** <https://content-dsxlocal.mybluemix.net/docs/content/local-dev/spss-modeler.html>

**Modeler documentation:**  
[https://www.ibm.com/support/knowledgecenter/en/SS3RA7\\_18.1.1/modeler.kc.doc/clementine/knowledge\\_center/product\\_landing.html](https://www.ibm.com/support/knowledgecenter/en/SS3RA7_18.1.1/modeler.kc.doc/clementine/knowledge_center/product_landing.html)

**Algorithms Guide:**  
<ftp://public.dhe.ibm.com/software/analytics/spss/documentation/modeler/18.1.1/en/AlgorithmsGuide.pdf>

**Compatibility Reports:**  
<http://publib.boulder.ibm.com/infocenter/prodguid/v1r0/clarity/prereqsForProduct.html>

Full or partial product name:



Search results:

- IBM SPSS Modeler
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- IBM SPSS Modeler for Data Science Experience
- IBM SPSS Modeler Gold