

# Cortana Intelligence Suite

Lab CIS007

Introduction  
to Azure  
Machine Learning



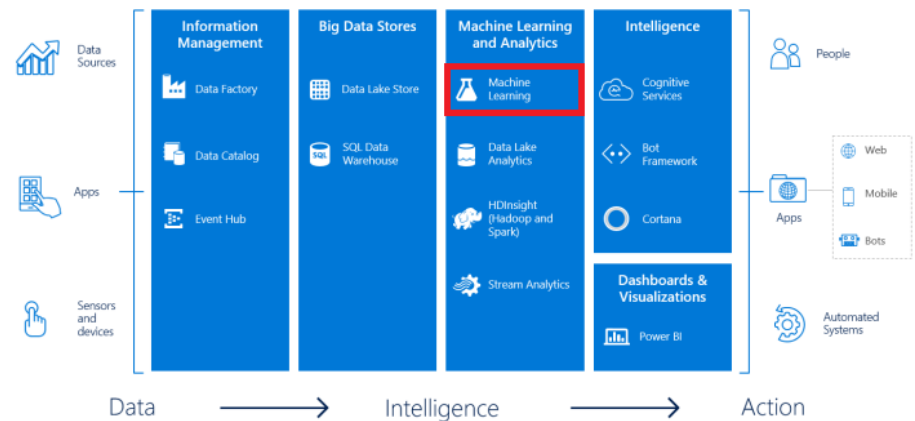
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# Overview

## Summary

This lab is intended to serve as an introduction to creating a predictive model with Azure Machine Learning.



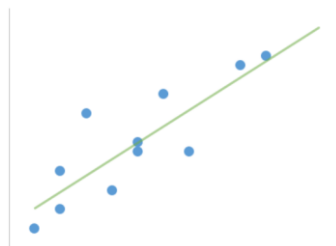
The lab will walk through accessing the ML Studio environment, exploring and visualizing data in Azure Machine Learning, and creating a simple predictive model.

## Business Case

Consumers often evaluate similar products using specific metrics of interest to them. In the auto industry, miles per gallon (MPG) always comes up as an important metric for consumers. How do manufacturers know what an acceptable MPG will be for the vehicle they are producing? Using advanced analytics, auto manufacturers can use vehicle attributes and MPG from similar automobiles already in the market in order to determine if the model rolling off the assembly line has a competitive MPG rating.

For this lab, you will be working with a dataset that includes various information about automobiles from the 1970s and early 1980s. The dataset includes attributes like miles per gallon (MPG), horsepower, acceleration, weight, etc.

The lab will use a linear regression algorithm to predict an acceptable MPG for an automobile. Linear regression is used to predict a single, numeric value based on one or many independent variables. It does this by fitting a representative line, or function, to a collection of input variables. This line/function can then be used to predict future values based on new input data.



## Learning Objectives

Upon completing this lab, you will have hands-on experience with the following functions and concepts related to Azure Machine Learning:

- Creating and logging in to a free Azure Machine Learning Workspace
- Creating, modifying, and saving an experiment with ML Studio
  - Running an experiment
  - Using sample datasets in an experiment
  - Browsing modules to use in an experiment
  - Using the search function to find modules for use in an experiment
  - Connecting modules in an experiment
  - Modifying and configuring properties of modules in an experiment
- Visualizing and exploring data in ML Studio
  - Exploring summary statistics about datasets and features
  - Visualizing features with scatterplots, boxplots, and histograms
  - Visualizing relationships between features with scatterplots and boxplots
  - Visualizing predictive model results
- Creating a basic predictive model in ML Studio
  - Splitting data into a training dataset and a test dataset
  - Training a Linear Regression model
  - Testing a trained model
  - Evaluating model performance

## Lab Requirements/Prerequisites

- A Microsoft account is required to access an Azure Machine Learning workspace. If you don't already have a Microsoft account, you can obtain one for free by following the link below:

<https://www.microsoft.com/en-us/account/default.aspx>

# Create a Workspace and Experiment

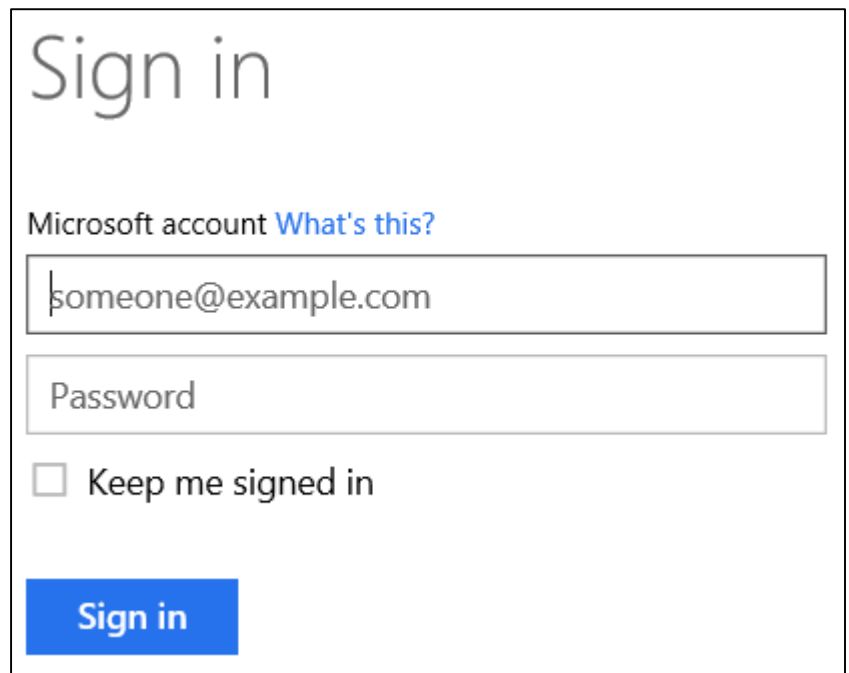
## Create/Access an Azure Machine Learning Workspace

To get started, you will need to create and log in to a free Azure Machine Learning workspace. A workspace is like an all-inclusive development environment with the tools to create, manage, and publish machine learning models.

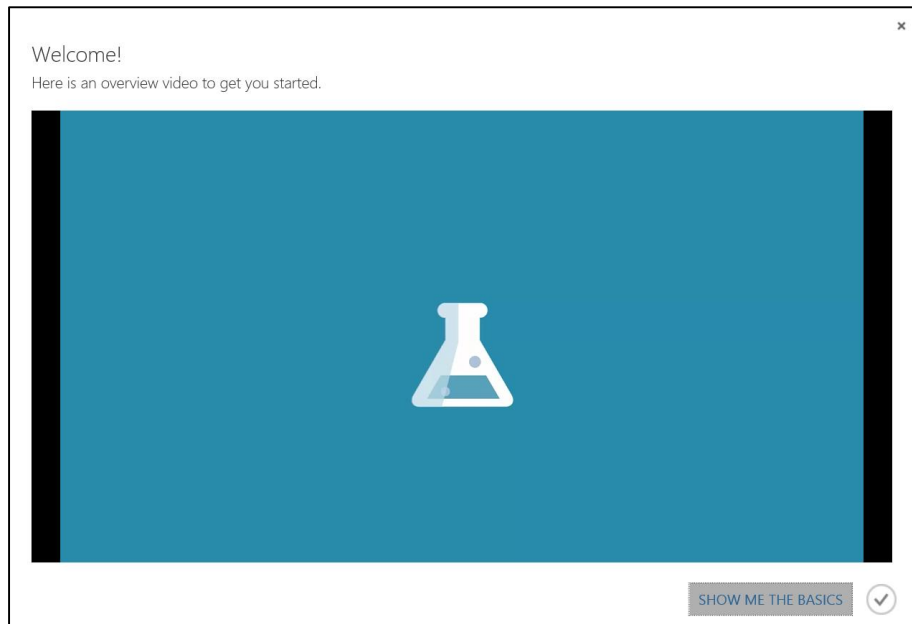
1. Open **Internet Explorer** or another preferred browser using the shortcut on the desktop taskbar.
2. Go to the **ML Studio** website by typing <http://studio.azureml.net> in the address bar.
3. Click **Sign In** on the top right corner of the web page.



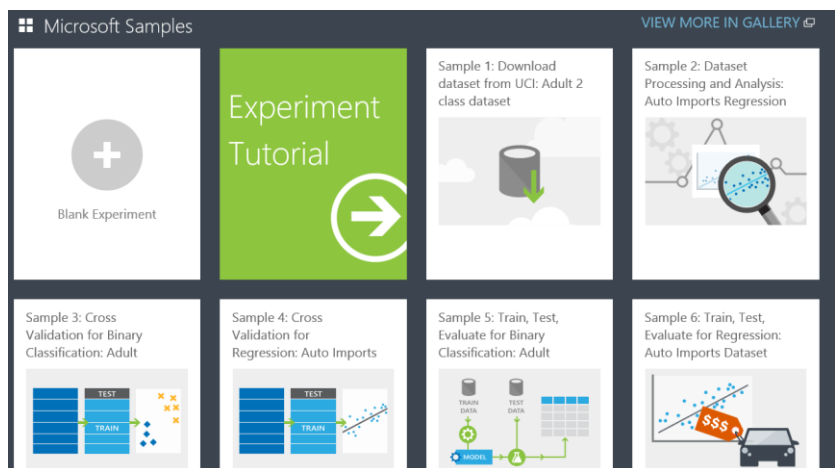
4. Enter the **email address** and **password** associated with your Microsoft ID, and click the **Sign In** button.

A screenshot of the ML Studio Sign in page. The page has a white background with a dark border. At the top, the text 'Sign in' is displayed in a large, light gray font. Below this, the text 'Microsoft account' is followed by a blue link 'What's this?'. There are two input fields: the first contains the email address 'someone@example.com' and the second is labeled 'Password'. Below the password field is a checkbox labeled 'Keep me signed in'. At the bottom, there is a blue button with the text 'Sign in' in white.

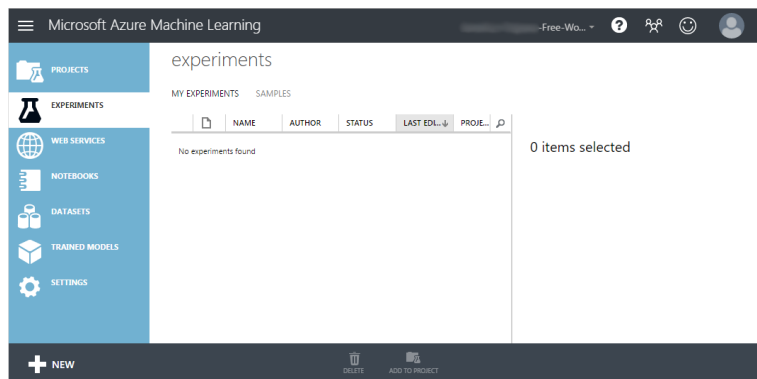
5. If upon logging in, a **Welcome** video is displayed (usually displays on the first login), click the **X** at the top right of the video to close it.



6. If the **Microsoft Samples** dialogue box is displayed (usually displays on the first login), go ahead and close it by choosing the **X** in the top right corner of the pane.



You are now logged into the free workspace associated with your Microsoft ID.



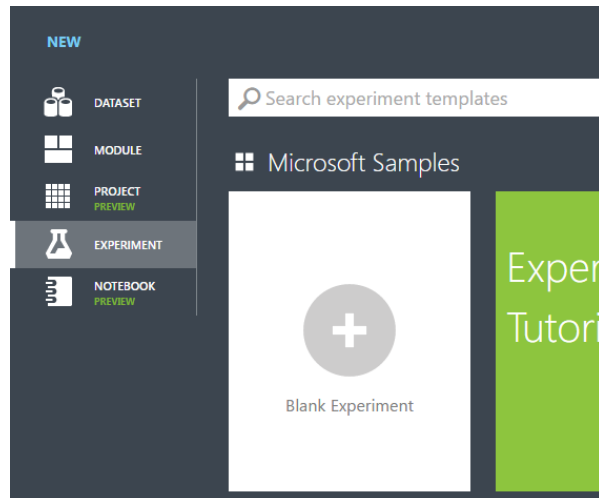
## Create a Blank Experiment

Next, we will create our first experiment. An experiment is a collection of data, tasks, and machine learning algorithms that make up a model.

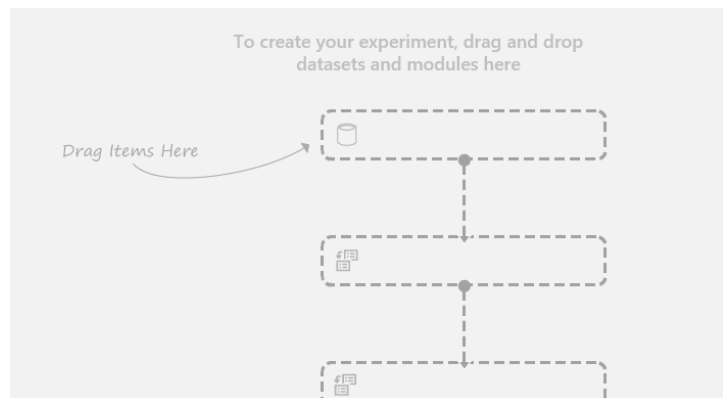
1. Click the **NEW** button in the bottom left corner of the page.



2. Make sure **EXPERIMENT** is highlighted in the **NEW** dialogue window, and click the **Blank Experiment** pane.



3. You are now in the **ML Studio**. Notice:
  - a. The *Canvas* in the center of the screen. This is where you will drag and drop modules and string them together to create a data flow for your experiment.



- b. The *Navigation icons* on the far left of the site allowing you to browse back to your *Workspace*



- c. The *Modules* pane down the left side of the Canvas are the individual components that make up your Experiment.

- › Saved Datasets
- › Trained Models
- › Transforms
- › Data Format Conversions
- › Data Input and Output
- › Data Transformation
- › Feature Selection
- › Machine Learning
- › OpenCV Library Modules
- › Python Language Modules
- › R Language Modules
- › Statistical Functions
- › Text Analytics
- › Web Service
- › Deprecated

- d. The *Properties* pane down the right side of the Canvas is where you will configure the properties of the different Modules used in your Experiment.

Properties

Project

Experiment Properties

STATUS CODE InDraft

Summary

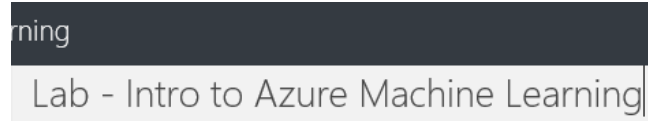
Enter a few sentences describing your experiment (up to 140 characters).

Description

Enter the detailed description for your experiment.



4. At the top of the **Canvas**, highlight and delete the text that reads **Experiment created on...**, and replace it with **Lab - Intro to Azure Machine Learning**.

A screenshot of the top of the Canvas interface. It shows a dark grey header bar with the text 'ning' on the left. Below this is a light grey bar containing the text 'Lab - Intro to Azure Machine Learning' with a vertical cursor at the end of the line.

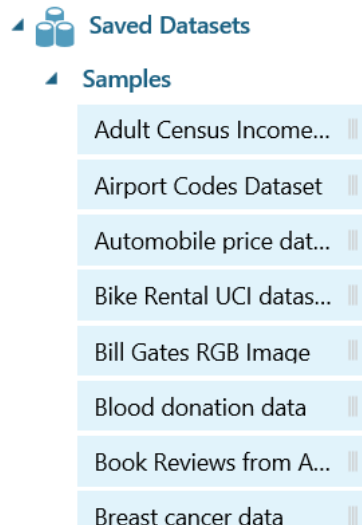
# Explore and Visualize Data

## Input Sample Data

Azure Machine Learning offers several ways to connect to and import data. For this lab, we will work with one of the sample datasets included with Azure Machine Learning.

1. On the **Modules** panel, click **Saved Datasets** and then **Samples**.

This expands all of the sample datasets included in ML Studio.



2. Scroll until you find **MPG data for various automobiles**.
3. Click on the **MPG** dataset and notice the description also shows up at the bottom of the **Properties** pane.
4. Click and drag the **MPG** dataset onto the **Canvas**. Notice that the Properties pane will reflect information about the dataset if you expand the arrow.

## Properties Project

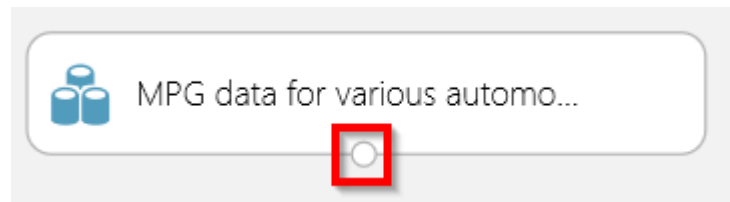
### MPG data for various automobi...

SUBMITTED BY	Microsoft C...
SIZE	17.4 KB
FORMAT	GenericCSV
CREATED ON	4/8/2015 6:...

[View dataset](#)

Notice at the bottom of the *MPG dataset* module on the Canvas that there is a small circle called a port. Ports on the top of modules are called input ports, and ports on the bottom

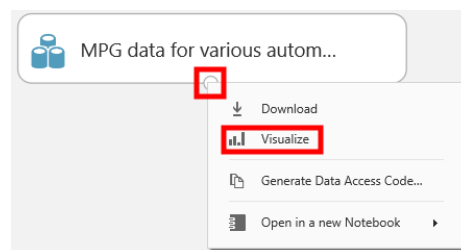
of modules are output ports. These ports are used to connect modules to one another and to provide a menu of additional options for the module.



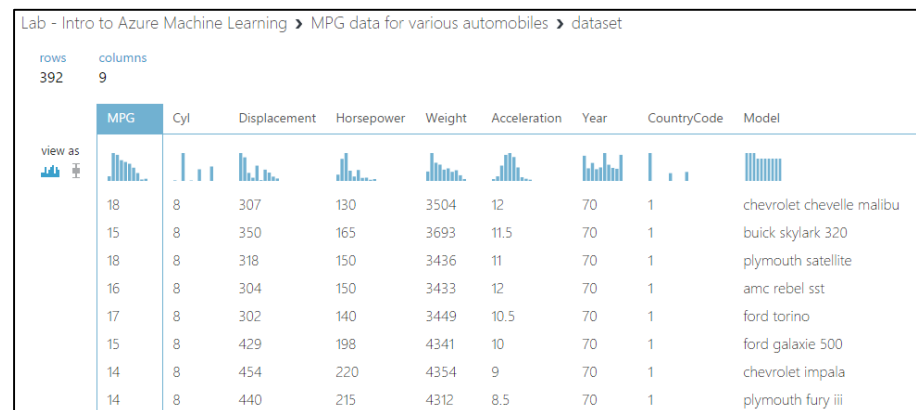
### Explore the Input Data

A common task in any advanced analytics workflow is to analyze and profile the data you are working with. The following set of steps highlights some of the ways we can explore and visualize the data we just imported.

1. Click the **output port** at the bottom of the **MPG** dataset module, and select **Visualize** from the menu that is displayed.



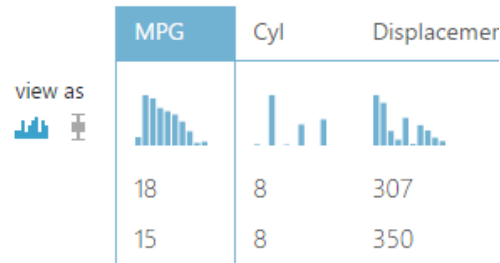
The resulting dialogue box provides the number of rows and columns in the dataset as well as the first 100 rows and first 100 columns in the dataset with a histogram for each column.



2. Click anywhere in the first column, **MPG**, to highlight the column.

rows  
392

columns  
9



Notice on the right side of the dialogue box, there is now information in the *Statistics* pane and *Visualizations* pane about MPG (you might need to use the horizontal scroll bar in the dialogue box to scroll all the way to the right if Statistics and Visualizations are not visible).

#### Statistics

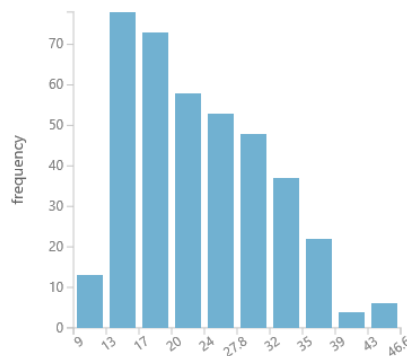
Mean	23.4459
Median	22.75
Min	9
Max	46.6
Standard Deviation	7.805
Unique Values	127
Missing Values	0
Feature Type	Numeric Feature

#### Visualizations

MPG

Histogram

compare to



- In the **Visualizations** pane, change the **compare to** dropdown box from **None** to **Horsepower**.

MPG

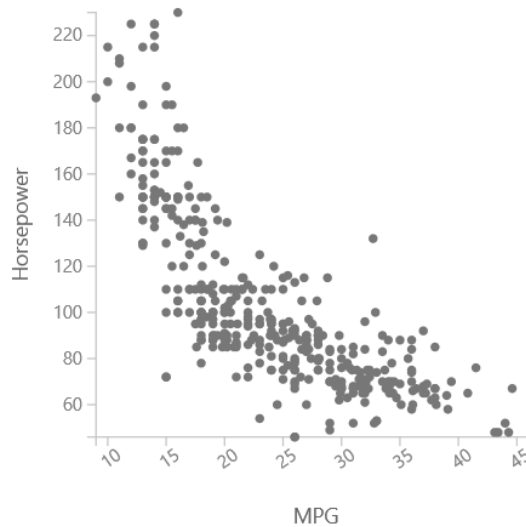
Histogram

compare to

None

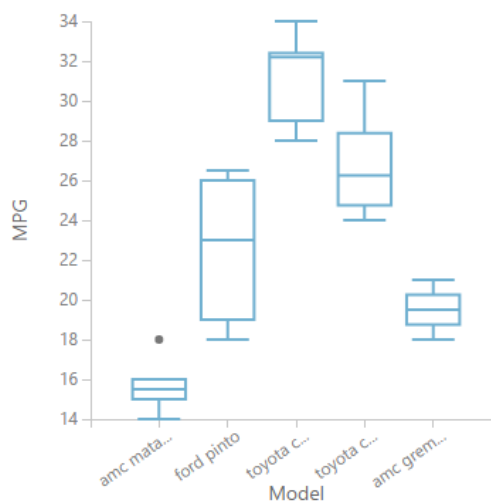


Notice the histogram changed to a *ScatterPlot* comparing MPG to Horsepower.



- Next, change the **compare to** dropdown option from **Horsepower** to **Model**.

Notice the resulting chart is now a *MultiboxPlot* with an MPG boxplot displayed for each of the values in the Model column.



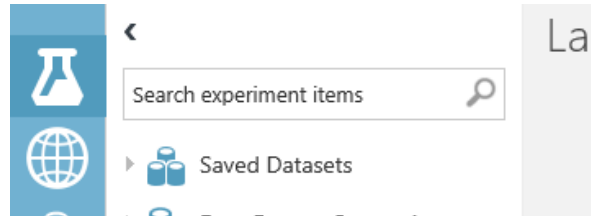
- Click the **X** in the top right corner of the **Visualize** dialogue box to return to the **Canvas**.

# Create a Simple Predictive Model

## Split Input Data into Train and Test Data Sets

Now that we have explored our data, we are ready to create a predictive model. The first thing we will do is split the original dataset into 2 datasets: one dataset will be used for training a model, and one will be used for testing our model. By testing the model on a separate (test) dataset, we can assess the accuracy of our training model.

1. In the search box at the top of the **Modules** pane, type the word **split**.



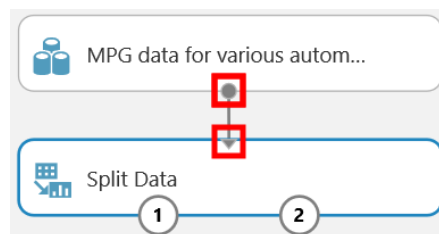
Notice the list of modules has been filtered to show only those relevant to the search term.

2. Click and drag the **Split Data** module onto the **Canvas** anywhere under the **MPG** dataset.

Notice the Split Data module has 1 input port and 2 output ports. The Properties pane displays values that can be modified for this module. There is also a description of the module at the bottom of the Properties pane with a (*more help...*) link. A page will open with more details about the module and its configurable properties when this link is clicked.



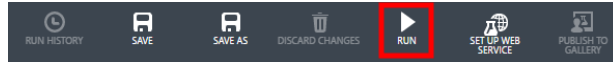
3. Click and drag the **output port** from the **MPG** dataset module to the **input port** of the **Split Data** module.



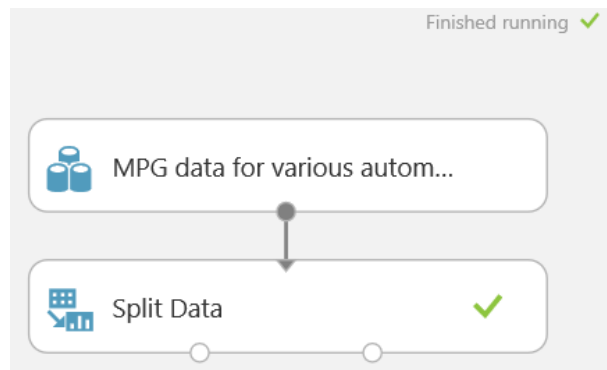
4. In the **Properties** pane, type **0.75** in the **Fraction of rows in the first output dataset** textbox.

This configures the module to split 75% of the input rows to the left output port, and 25% of the input rows to the right output port. This means that 75% of the data will be used to train the model, while the remaining 25% will be used to score it for accuracy.

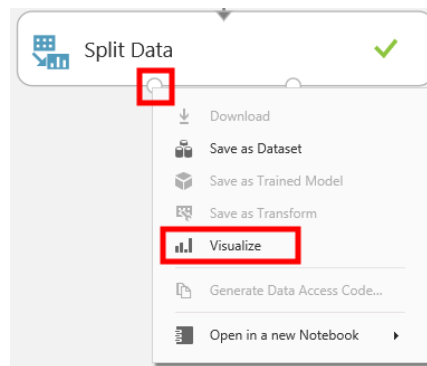
5. Click **RUN** at the bottom of the **Canvas**.



The experiment will now execute each module in order, beginning with the first module in the workflow. When the experiment is done executing, the words *Finished running* will display in the top right corner of the Canvas. Notice the Split Data module has a *green check mark* indicating it completed successfully.



6. Click the **left output port** on the **Split Data** module, and select **Visualize** from the menu that is displayed.



- 7.

Notice only 294 of the original 392 rows (75%) have been routed to the left output port. The remaining 98 rows (25%) have been routed to the right output port.

8. Click the **X** in the top right corner to close the **Visualize** dialogue box.

### Train a Predictive Model

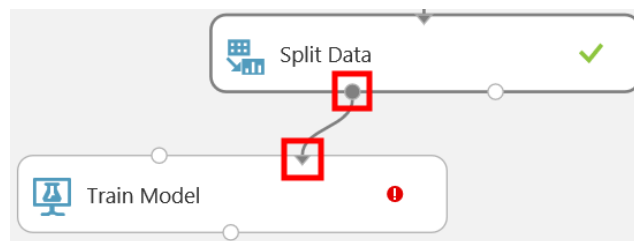
Next, you will use a common Linear Regression algorithm to train a model that will predict an automobile's MPG. Linear Regression attempts to find a linear relationship between numeric variables.

1. Type **train** in the search box at the top of the **Modules** pane.
2. Find the **Train Model** module, and click and drag it onto the **Canvas** below the **Split Data** module.

#### Train



3. Connect the **left output port** from the **Split Data** module to the **right input port** on the **Train Model** module

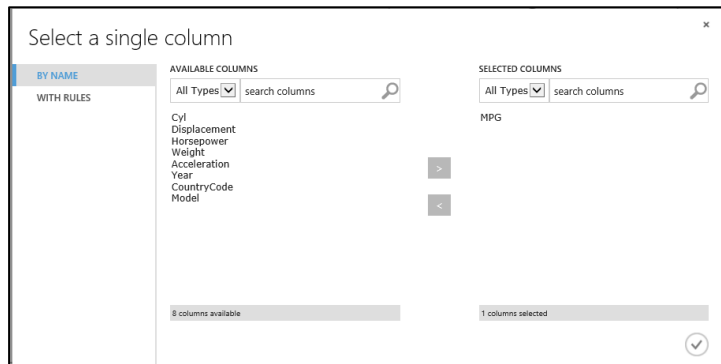



4. In the **Properties pane**, click the **Launch column selector** button.

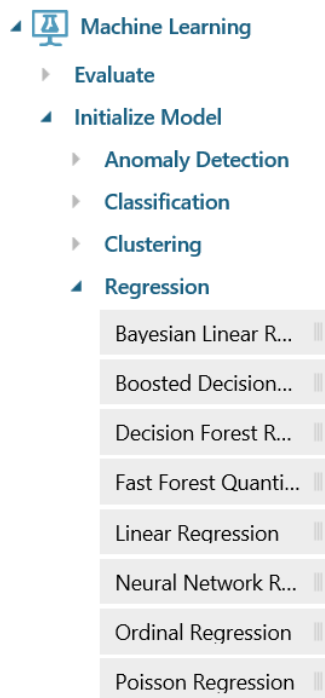
This launches the *Select Column* dialogue box. Here, we will select the column we want the model to predict.

5. In the **Available Columns** list, select **MPG**.
6. Click the **>** button to move **MPG** to the **Selected Columns** list.

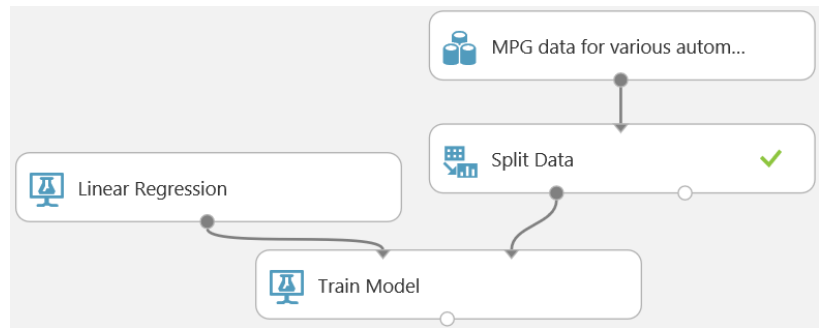




7. Click the  button to save the selection and close the dialogue box.
8. Clear the search box in the **Modules** pane.
9. In the **Modules** pane, find and click to expand **Machine Learning**, and then click **Initialize Model**, and then click **Regression**.



10. Click and drag the **Linear Regression** module onto the **Canvas** just above and to the left of the **Train Model** module.
11. Connect the **output port** of the **Linear Regression** module to the **left input port** of the **Train Model** module.



You might notice there are several parameters that can be modified in the Properties pane for the Linear Regression module. For this lab, we will use the defaults.

12. Click **RUN** at the bottom of the **Canvas** to run the experiment and train the model.

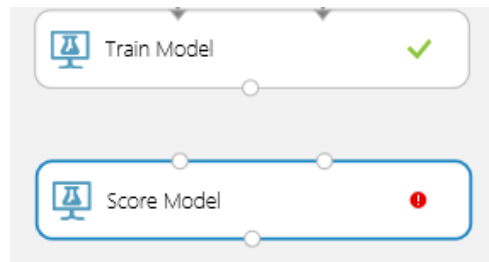


The model will be trained to predict the MPG column using the other fields in the dataset with the Linear Regression algorithm.

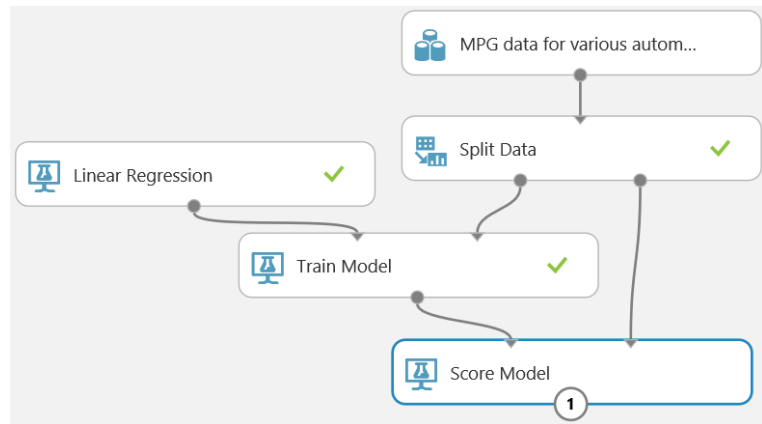
### Test the Predictive Model

Next, we will use the test dataset we created to test our newly trained model. This will be accomplished using our new model to predict the MPG for each row in the test dataset.

1. In the search box at the top of the **Modules** pane, type the word **score**.
2. Find the **Score Model** module, and click and drag it onto the **Canvas** under the **Train Model** module.



3. Connect the **output port** on the **Train Model** module to the **left input port** on the **Score Model** module.
4. Connect the **right output port** on the **Split Data** module to the **right input port** on the **Score Model** module.



- Click **RUN** at the bottom of the **Canvas** to run the experiment and score the test dataset with the trained Linear Regression model.



- After the experiment has finished running, click the **output port** on the **Score Model** module and select **Visualize** from the displayed menu.
- In the list of columns, scroll to the right until **Scored Labels** is visible, and click **Scored Labels** to select it.

Acceleration	Year	CountryCode	Model	Scored Labels
19.6	82	1	chevrolet cavalier	25.078859
16.2	82	1	pontiac j2000 se hatchback	27.133106
15.9	77	1	dodge colt m/m	27.031567
14.5	71	2	fiat 124b	27.676992

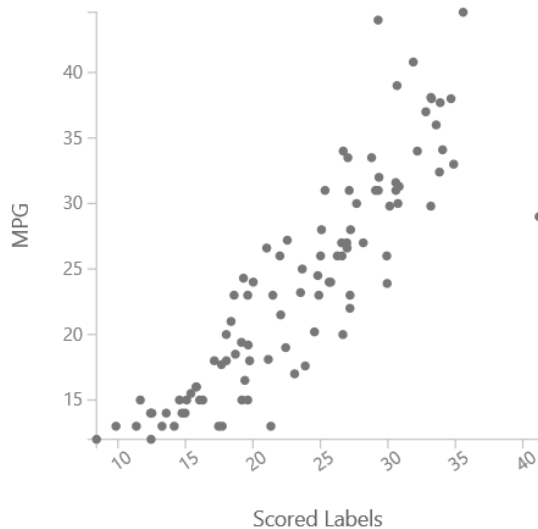
The *Scored Labels* column represents the predicted MPG for each row in the test dataset. Notice the *Statistics* pane and histogram in the *Visualizations* pane on the right side of the *Visualize* dialogue box.

- In the **Visualizations** pane, change the **compare to** dropdown option to **MPG**.

### Scored Labels

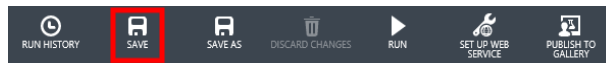
#### ScatterPlot

compare to



The resulting *ScatterPlot* compares the Scored Labels (predicted MPG) with the actual MPG for each row in the test dataset.

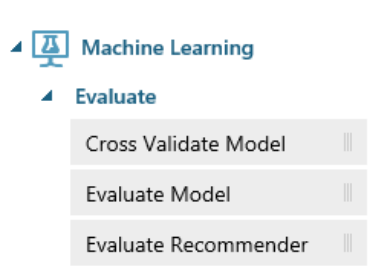
9. Click the **X** in the top right corner to close the **Visualize** dialogue box.
10. Click **SAVE** to save the experiment.



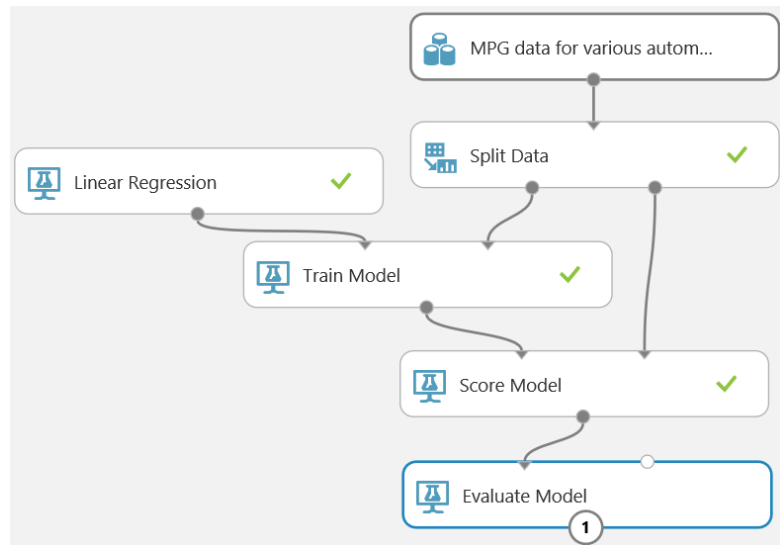
### Evaluate the Test Results

Finally, we will evaluate how well the model performed against the test dataset using a set of standard metrics for measuring regression model performance.

1. In the search box at the top of the **Modules** pane, type the word **evaluate**.
2. Find the **Evaluate Model** module, and click and drag it onto the **Canvas** below the **Score Model** module.



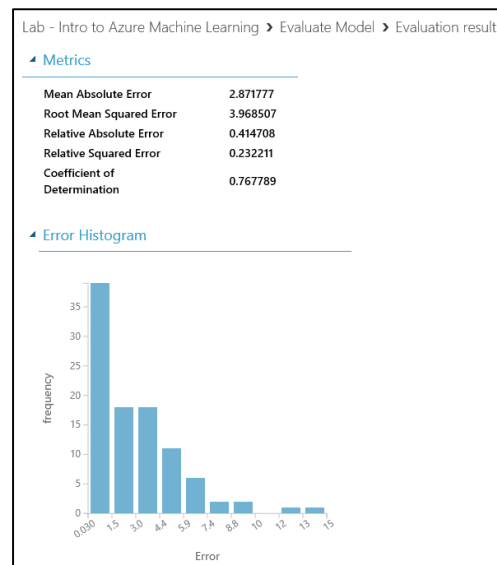
3. Connect the **output port** on the **Score Model** module to the **left input port** on the **Evaluate Model** module.



4. Click **RUN** to run the experiment.



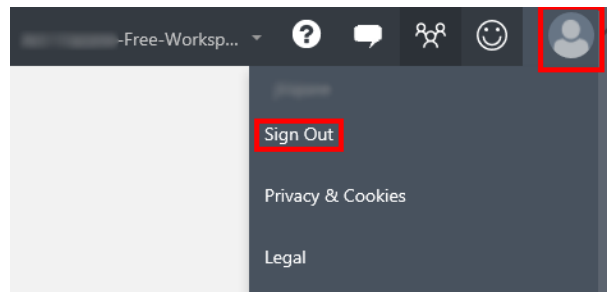
5. When the experiment has finished running, click the **output port** on the **Evaluate Model** module and select **Visualize** from the displayed menu.



The columns and values in the *Visualize* dialogue box represent common metrics for evaluating the performance of a Linear Regression model. The metrics are calculated using the results of the *Score Model* module. Many of the metrics are based on the Error, which is the difference between the Scored Labels (predicted value) and the actual values.

At this point, you can assess whether or not your model performs at a satisfactory level. If not, you could go back and tweak parameters, add new features, try a different machine learning algorithm, etc., in an attempt to induce more acceptable results from the model.

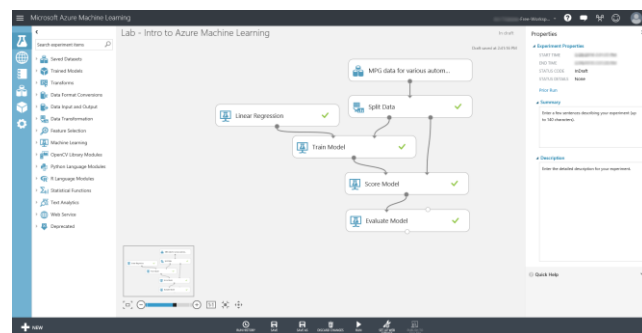
6. Click the **X** in the top right corner to close the **Visualize** dialogue box.
7. Click **SAVE** to save your experiment
8. Sign out of your workspace by clicking the **profile picture** at the top right of the page and selecting **Sign Out** from the displayed menu.



## Conclusion

This concludes the *Intro to Azure Machine Learning* lab. To recap, you have successfully created and logged into an Azure Machine Learning workspace, explored and visualized data in ML Studio, and produced and saved a simple Linear Regression model that predicts automobile MPG.

You now have a model that can be used by an auto manufacturer to predict an acceptable MPG for any new automobile coming off the assembly line. As a next step, you could upload new data to be scored by your model, or you could even publish your model as a web service. A web service gives you the ability to pass individual rows in and get predicted values (MPG) out.



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