Azure Machine Learning



Hello!

Il cielo è Azure sopra Berlino team

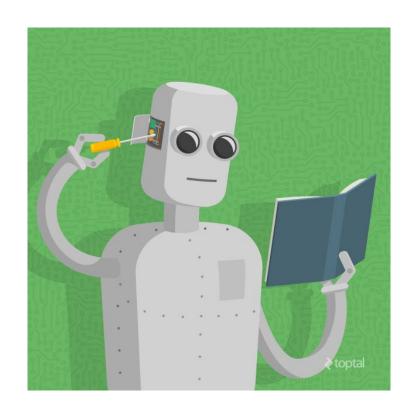
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Machine Learning

- What is Big Data?
- What is Machine Learning?
- Uses of Machine Learning?
- Why Machine Learning?
- Who uses it?



What is big data?



- What is Big Data?
 - Structured
 - Unstructured
- From a variety of sources
 - Commercial transactions
 - Social media
 - Publicly available sources
 - Sensors
 - Business statistics
- How to analyze this data?



What is machine learning?

- Examine LARGE amounts of data
 - Find patterns. Build models.
- Automatic improvement of the algorithms
 - Iterative approach.
 - Multiple passes so the machine learns.
- Predictions







Uses of machine learning?

- Classification
 - Supervised.
 - e.g. spam filter
- Regression
 - Supervised.
 - Estimate relationship between continuous variables.
 - e.g. car market price from specs
- Clustering
 - Unsupervised.
 - e.g. identify communities in social networks



Why machine learning?

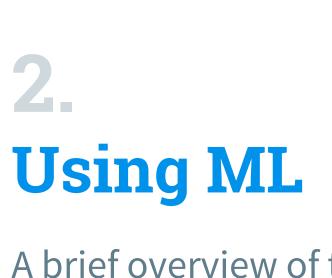
- Growing volumes and varieties of available data
 - Processing this data manually would be impossible.
- Cheaper computational processing and storage
- Competitive advantage
 - Companies get huge benefits by analyzing data from the markets.



Who uses it?

- Financial institutions
 - e.g. recognize and prevent frauds.
- O Governments
 - e.g. increase efficiency and service.
- Medicine and science
 - e.g. dna sequencing, patients wearable sensors.
- Marketing and sales
 - e.g. dna sequencing, patients wearable sensors.
- You name it!

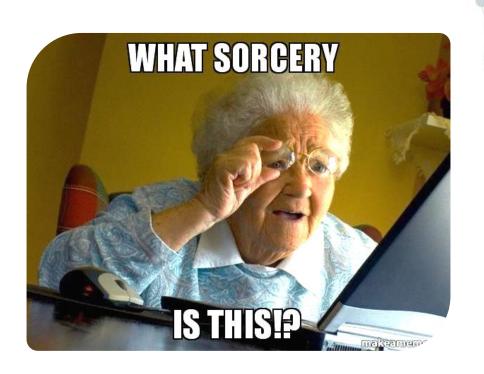




A brief overview of the current tools to harness the power of ML

ML is an incredibly powerful set of...

- Algorithms
- Tools
- Techniques
- ...
- Magic spells?!



Back in the ol' days...

To use ML, you'd have to implement the algorithms yourself:

- prototype in some kind of friendlier language (like Matlab/Octave);
- then implement it in a real language (like C++) for speed and efficiency.

Back in the ol' days...

In-depth knowledge of ML techniques and algorithms was *required*.

Huge barrier to adoption.



ML was used only in very big, very serious applications (that could afford and justify the *overhead*).



Keep it simple, so you'll keep doing it.

Tools to the rescue!

Libraries and frameworks lower the initial effort required to get a working prototype.



ML libraries and frameworks

- Exist for practically any widely used programming language.
- Encapsulate most widely used algorithms, abstracting away low-level details.
- Can even offer ad-hoc solutions for greater speed/efficiency/reliability (e.g. distributed computation).

ML libraries and frameworks The celebrities:















MLaaS

Cloud Computing approach gives us Machine Learning as a Service.



ML as a Service

Outsourcing ML services:

- Incredibly low barrier to adoption.
- Massive scalability.
- It just works!





ML as a Service - The celebrities:

- Google Prediction APIs
- Amazon AWS ML

- Microsoft Azure ML
 - Allows users to create and train models, then turn them into ready-to-be-consumed APIs.
 All through a beautifully intuitive web interface.

3. Azure Machine Learning Studio

Azure's solution to make your own experiments

What is Azure Machine Learning Studio?

- Web-based workspace.
- O Drag-and-drop tool.
- Collaborative environment.
- Where data science, cloud resources, and your data meet.

With Azure ML, predictive analytics solutions are...



Easy to build.

Easy to deploy.

Easy to share.

Ease of use!

ML can do amazing things... But they could be even more amazing if accessible to all!





Setup

All you need is a **web browser!** Go to Azure ML website and choose:

- Free workspace: start using all the features of Studio immediately, no credit card required!
- Enterprise workspace: add extra storage and few additional web services features (\$10/month).

Then, start working on your data from anywhere!

Build

Creating a predictive model with Azure ML is as easy as ...



... playing with LEGO®!

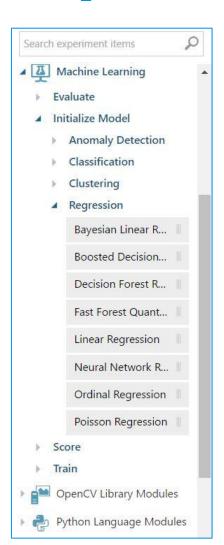
Build - main features

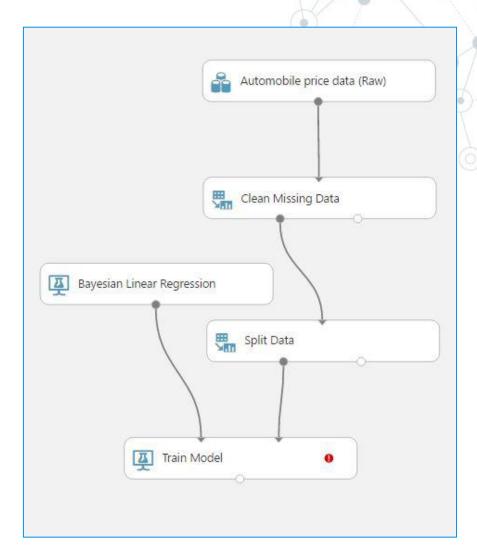
To help you building your **training experiment** (model) from scratch, Studio provides:

- Interactive, intuitive visual workspace.
- Drag-and-drop interaction to connect modules with each other. For instance:
 - ready-to-use **datasets**.
 - ready-to-use standard ML algorithms.
 - your special sauce (cooked in Python or R).
 - 0 ...
- Huge set of samples and templates.

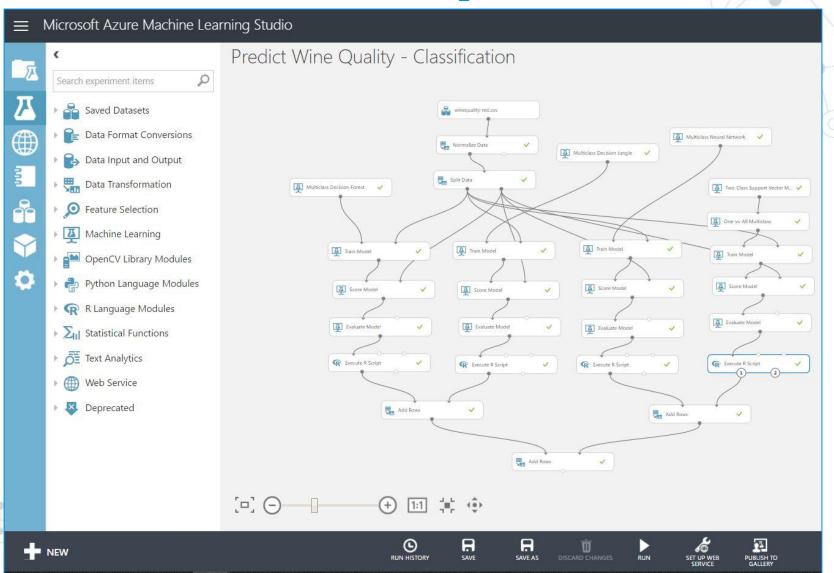
Build - example







Build - advanced example



Build - additional features

Besides creating experiments, Studio allows you to:

- upload your own datasets.
- o create web services. (!!!)
- store and reuse your trained models.
- o create Jupyter notebooks.
- save your account settings.
- ocollect all previous objects into a single **project**.

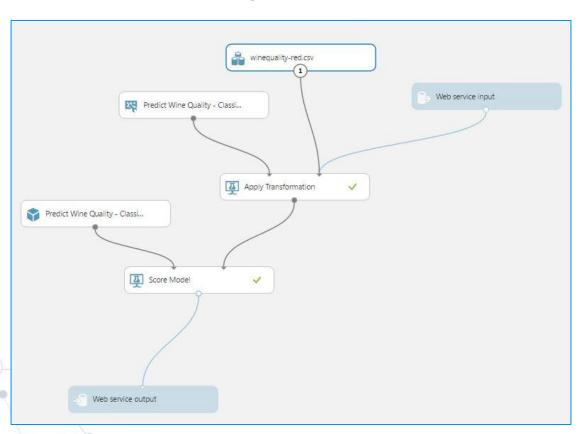
Deploy

Once your model is ready, deploy it as a **web service** in few steps:

- right from Studio, click on "Setup WS".
- wait for your **predictive experiment** to be created.
- click on "Deploy WS".
- wait for your web service to be deployed.
- enjoy!

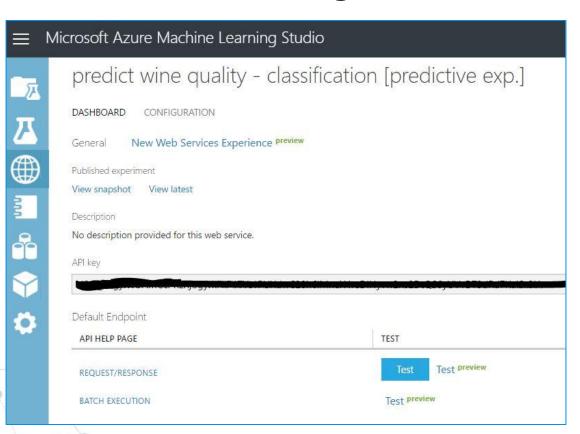
Deploy - predictive experiment

The original experiment is "translated" and the model is used to predict results.



Deploy - web service

To call your new web service, just follow the instructions about building the **POST request**.



Share

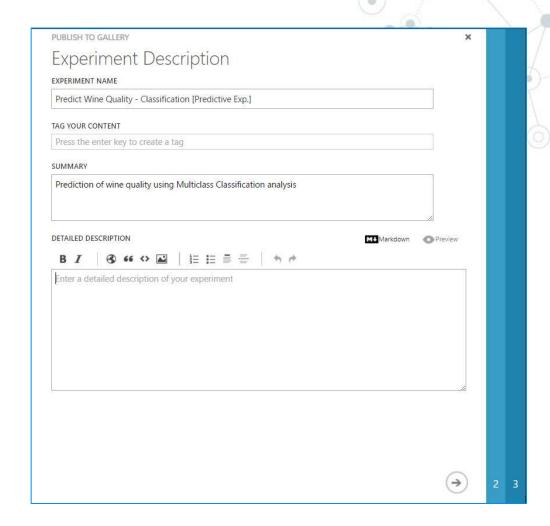
Your brand new experiment is ready to be shared in the community. Remember,

ML accessible for all!

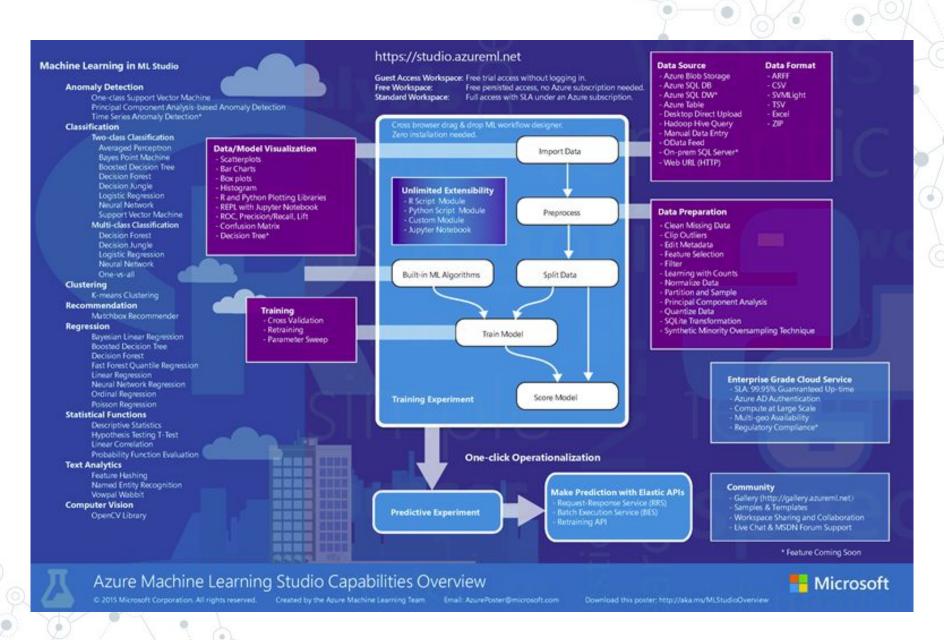
Upload it on **Cortana Intelligence Gallery**, where data scientists and developers share solutions.

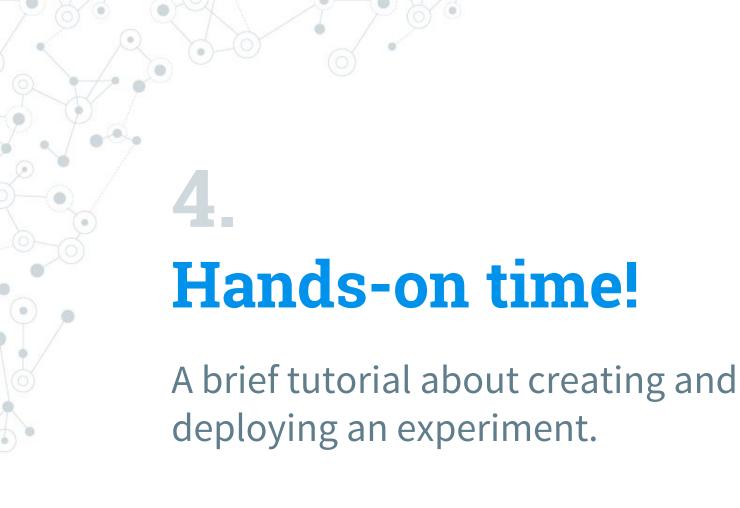
Share - gallery

You can publish your work directly from the Studio.
Just follow the instructions and describe what you have done!









Microsoft Azure Machine Learning Studio

- Go to Microsoft Azure Machine Learning Studio.
- In order to use the framework we need a Microsoft account:
 - A. I already have one of them→ just "Sign in"
 - B. I do not have any of them → must "Sign Up"

Welcome to Azure Machine Learning

Try it for free

No Azure subscription? No credit card? No problem! Choose anonymous Guest Access, or sign in with your work or school account, or a Microsoft account.



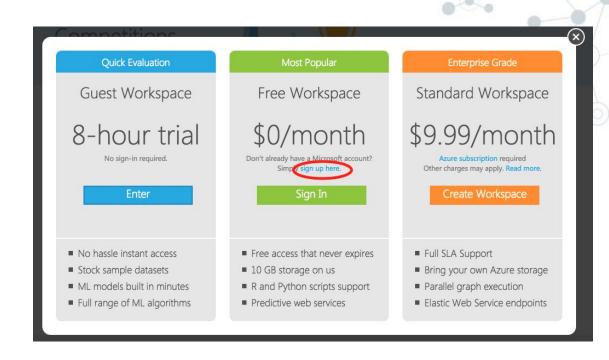
Already an Azure ML User? Sign in here



Sign up

Select "Free Workspace"

- Free access
- 10GB Storage
- R and Python scripts support
- Predictive web services





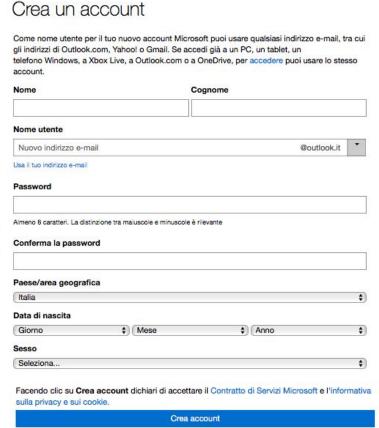
Create an account

- 1. Fill the form
- 2. Click on create an account
- 3. Verify your email



Verifica il tuo indirizzo e-mail

Per completare la configurazione di questo account Microsoft, dobbiamo verificare che questo indirizzo e-mail sia il tuo.



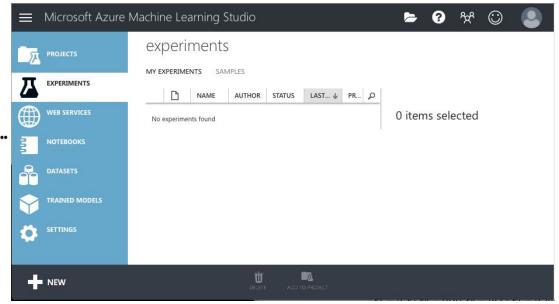
Microsoft

Sign in

Type the account you want to use and log in in the free workspace.



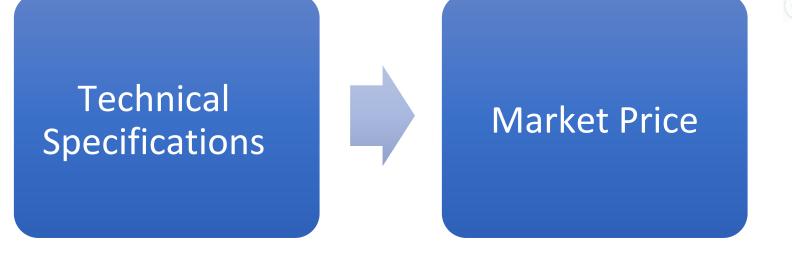
Microsoft Azure



Five steps to create an experiment

- Create a model
 - Get data
 - Prepare the data
 - Define features
- Train the model
 - Choose and apply a learning algorithm
- Score and test the model
 - Predict new automobile prices

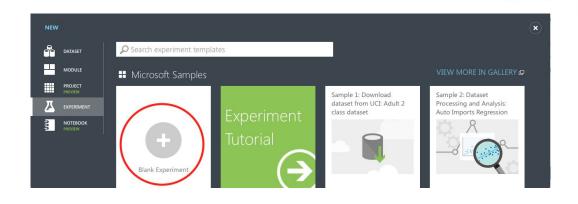
Automobile price prediction

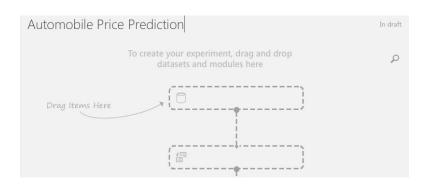




Create a blank experiment







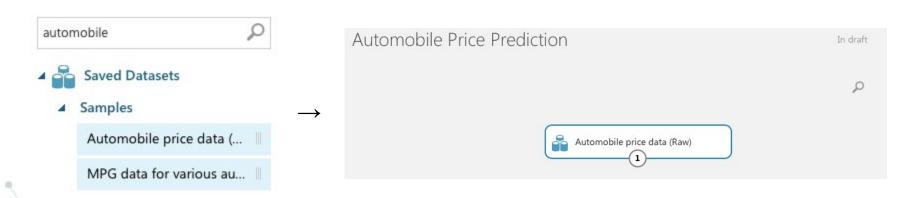
1. Get Data

- Use data in the existing sample datasets
- Create your own dataset by NEW dataset
- Import data: Load data from sources such as the Web, Azure SQL database, Azure table, Hive table, or Windows Azure BLOB storage. Formerly known as Reader

Using Azure saved dataset

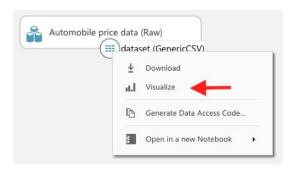
In the search bar, look for automobile

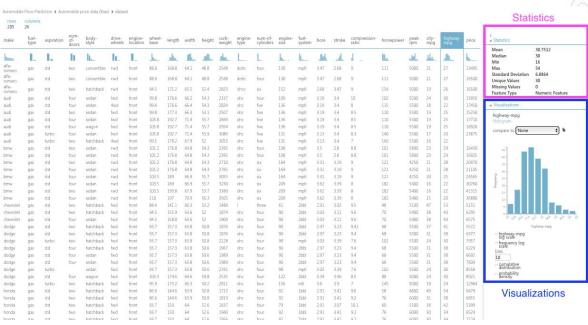
Drag and drop the dataset in the dashboard



Visualize the Data

- Selecting one column, some statistics are shown
- Given the variables for a specific automobile, we're going to try to predict the price (last column)





2. Prepare the data

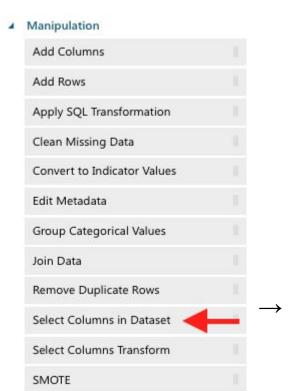
 This menu can be used to transform raw data to the input of the next modules



- Filter
- Learning with Counts
- Manipulation
- Sample and Split
- Scale and Reduce



Preprocess automobile dataset

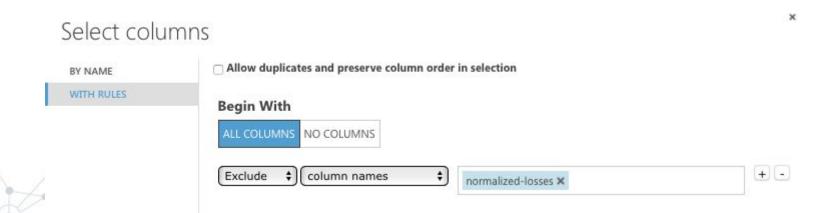


- 1. Clean missing values present in the columns of various rows so the model can analyze the data correctly.
- 2. Do not consider some columns.



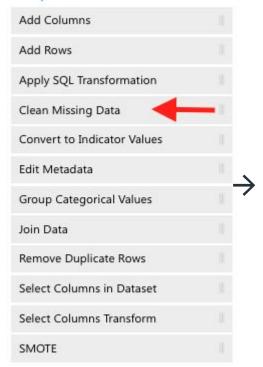
Clean missing data: remove column

- Olick on Launch column selector
- On the left, click With rules
- Under Begin With, click All columns.
- Select Exclude and column names,
- Click inside the text box and select normalized-losses.

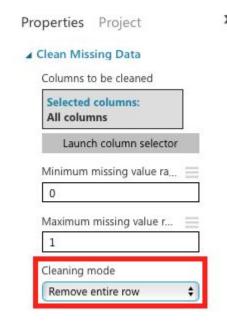


Clean missing data: remove row

▲ Manipulation



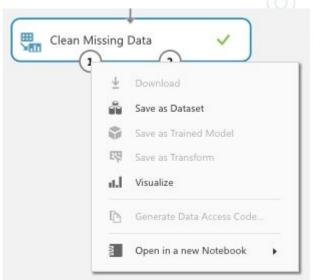




Run the experiment and visualize processed data

- Save the experiment
- Run it
- Visualize data output from Clean Missing Data
- Check differences





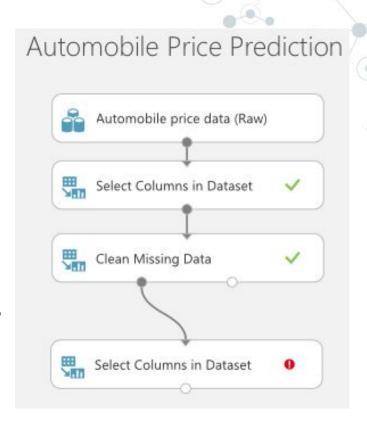


3. Define features

- Features: individual measurable properties of something you're interested in.
- Finding a good set of features for creating a predictive model requires experimentation and knowledge about the problem you want to solve.
- (In our example each row represents one automobile, and each column is a feature of that automobile)

Feature selection

- As before, drag Select columns in Dataset
- Connect Clean Missing Data to the module just added
- Olick on Launch column selector
- On the left, click **With rules**
- Under Begin With, click No columns.
- Select Include and column names,
- Click inside the text box and select "make", "body-style", "wheel-base", "engine-size", "horsepower", "peak-rpm", "highway-mpg", "price"





4. Choose and apply a learning algorithm

- Classification: predicts an answer from a defined set of categories
- Regression: predicts a number.
- (Because we want to predict price, which is a number, we'll use a regression algorithm)

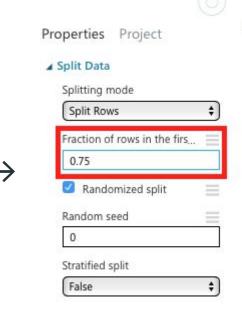
Build predictive model



Split data into train set and test set

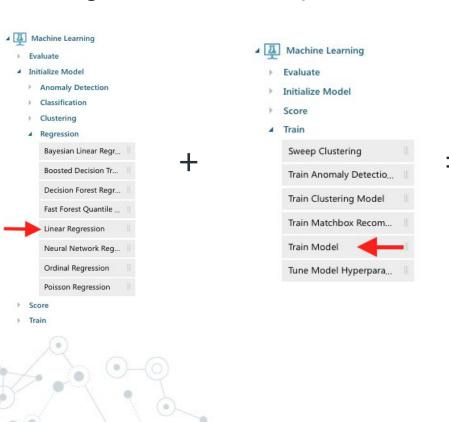


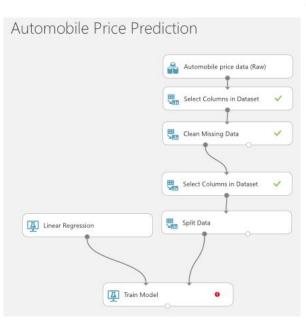




Learning algorithm selection

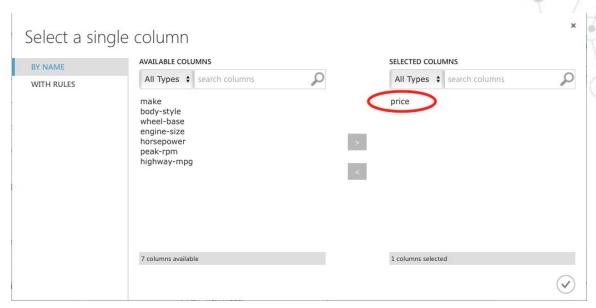
Connect the "Train Model" module to both the "Linear Regression" and "Split Data" modules





Train a specific feature

- Click the Train Model module
- Click Launch column selector in the Properties pane
- O Click By Name
- Select the **price** column.
- This is the value that our model is going to predict.



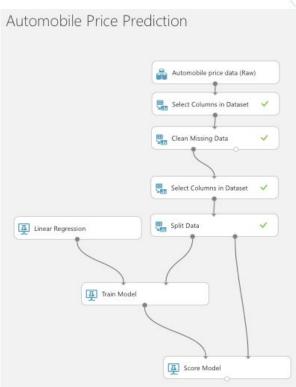


5. Predict new automobile prices

 75 percent of our data used to train the model using

 25 percent of the data to score the model functions.





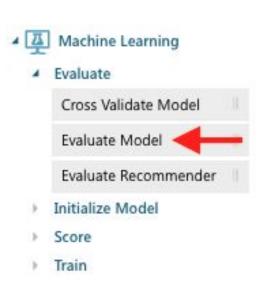
Output of the score module

Automobile Price Prediction > Score Model > Scored dataset

Predicted values for price and its probability.

Real Price Predicted Price wheelhighway engine body-style horsepower mpg Statistics view as Mean 12437.776 Median 10208.7085 5446.8479 29 10286.204819 subaru sedan 108 4800 11259 34960.6439 93.7 92 5500 38 5446.847864 mitsuhishi hatchback Standard Deviation 7323.458 Unique Values 46 dodge hatchback 93.7 90 5500 38 6229 6344.800711 Missing Values 5528.302953 hatchback 92 6000 38 6855 Feature Type Numeric Score 13498 476233 alfa-romero 88.6 130 27 16500 convertible 28 16515 16097.608038 volvo 104.3 141 114 5400 Visualizations wagon hatchback 119 5000 29 11048 8315.257218 Scored Labels 6630.154608 dodge hatchback 93.7 90 5500 41 5572 29 16430 19913.408695 sedan 108 101 5800 compare to None 93.7 92 68 5500 41 5389 5732.201761 mitsubishi hatchback 22 41315 30548.819502 bmw sedan 209 182 5400 113 258 176 4750 19 35550 30863.486076 jaguar sedan 18-38 5806.676601 plymouth hatchback 93.7 90 6229 16-16558 17388.014192 toyota hatchback 102.9 171 5200 24 14-14489 13094.447938 mitsubishi hatchback 95.9 156 145 5000 24 12-6092.030497 plymouth hatchback 93.7 90 68 41 10volkswagen 97 4800 46 8344.693482 dodge hatchback 93.7 98 102 5500 30 8258.383335 mercedes-115.6 234 4750 18 34184 34960.643871 benz alfa-romero hatchback 94.5 152 154 5000 26 16500 14329.816126

Results evaluation





(Final Experiment)

Metrics

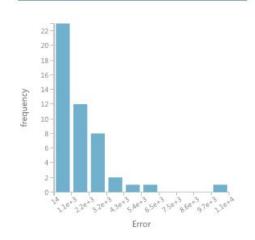
- Mean Absolute Error (MAE): The average of absolute errors (an error is the difference between the predicted value and the actual value).
- Root Mean Squared Error (RMSE): The square root of the average of squared errors of predictions made on the test dataset.
- Relative Absolute Error: The average of absolute errors relative to the absolute difference between actual values and the average of all actual values.
- Relative Squared Error: The average of squared errors relative to the squared difference between the actual values and the average of all actual values.
- Coefficient of Determination: Also known as the R squared value, this is a statistical metric indicating how well a model fits the data.

Automobile Price Prediction > Evaluate Model > Evaluation results

■ Metrics

Mean Absolute Error	1656.147651
Root Mean Squared Error	2456.983209
Relative Absolute Error	0.276606
Relative Squared Error	0.089608
Coefficient of Determination	0.910392

Error Histogram



How a metric should be

- For each of the error statistics, smaller is better.
- A smaller value indicates that the predictions more closely match the actual values.
- For Coefficient of Determination, the closer its value is to one (1.0), the better the predictions.

Iterate to improve the model

- Change the features you use in your prediction
- Modify the properties of the Linear Regression algorithm
- Try a different algorithm altogether
- Add multiple machine learning algorithms to your experiment at one time
- Compare two of them by using the EvaluateModel module

6. Deploy an Azure Machine Learning web service

- Satisfied with your model????
- You can deploy it as a web service!
- Use the WebService to predict automobile prices by using new data...

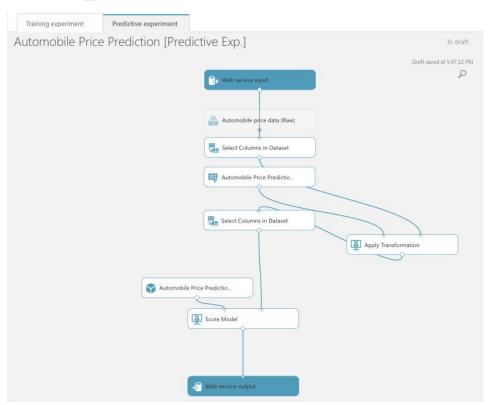


Convert the training experiment to a predictive experiment

- By converting to a predictive experiment, you're getting your trained model ready to be deployed as a scoring web service.
- Users of the web service can send input data to your model and your model will send back the prediction results.
- As you convert to a predictive experiment, keep in mind how you expect your model to be used by others.



Predictive experiment

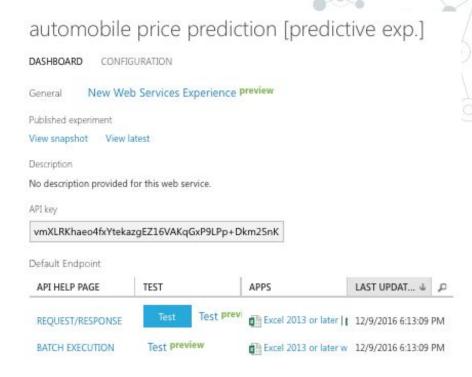




Deploy the predictive experiment as a New web service

- Click Run
- Click Deploy Web Service
- Select Deploy Web Service New.
- The deployment page of the Machine Learning Web Service portal opens.

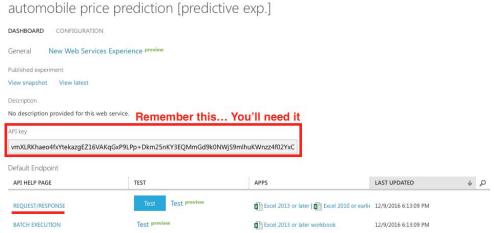




Test your Web Service with a Python Program



 request/response page contains Request Response API Documentation, with a starter Python program (that must be modified) to call the web service



Available material



https://github.com/giacomolanciano/Azure-Machine-Learning-tutorial



http://www.slideshare.net/GiacomoLanciano/azure-machine-learning-tutorial



Thanks!

Any questions?

