

Machine Learning Simplified for Developers with ML.NET



ML.NET

Virtual ML .NET Community Conference

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Brisbane AI User Group



.NET Core and EF Core dev by day

Cognitive Services and ML.NET enthusiast at night



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[Home page](#)[Face API](#) ^[Explore](#)[Detect/Identify](#)[Text API \(beta\)](#)[Speech API \(alpha\)](#)[Form Recognizer \(alpha\)](#)[HOME](#) > [FACE](#) > [V1.0](#) > [DETECT](#)

Detect

Profiles
PoC

+

✎

1. Get image

Image URL

File Upload

Web cam

2. Detect faces

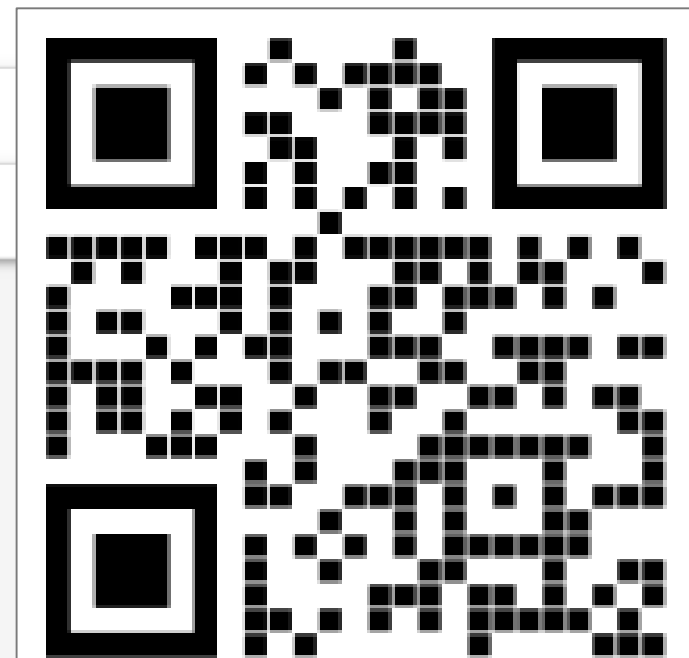
Detection model
v02 (recommended) ▼[More details](#)Recognition model (based on the g
v01 (group default) ▼[More details](#)[Detect \(1 TR\)](#)



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ML.NET Model Builder

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Investment

COFFEE 7 Melbourne



Food & Drink

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Education



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Azure ML



TensorFlow


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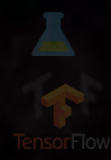
 Investment

 Food & Drink

 Education

ONE DOES NOT SIMPLY

LEARN MACHINE LEARNING





IF TIME IS INFINITE

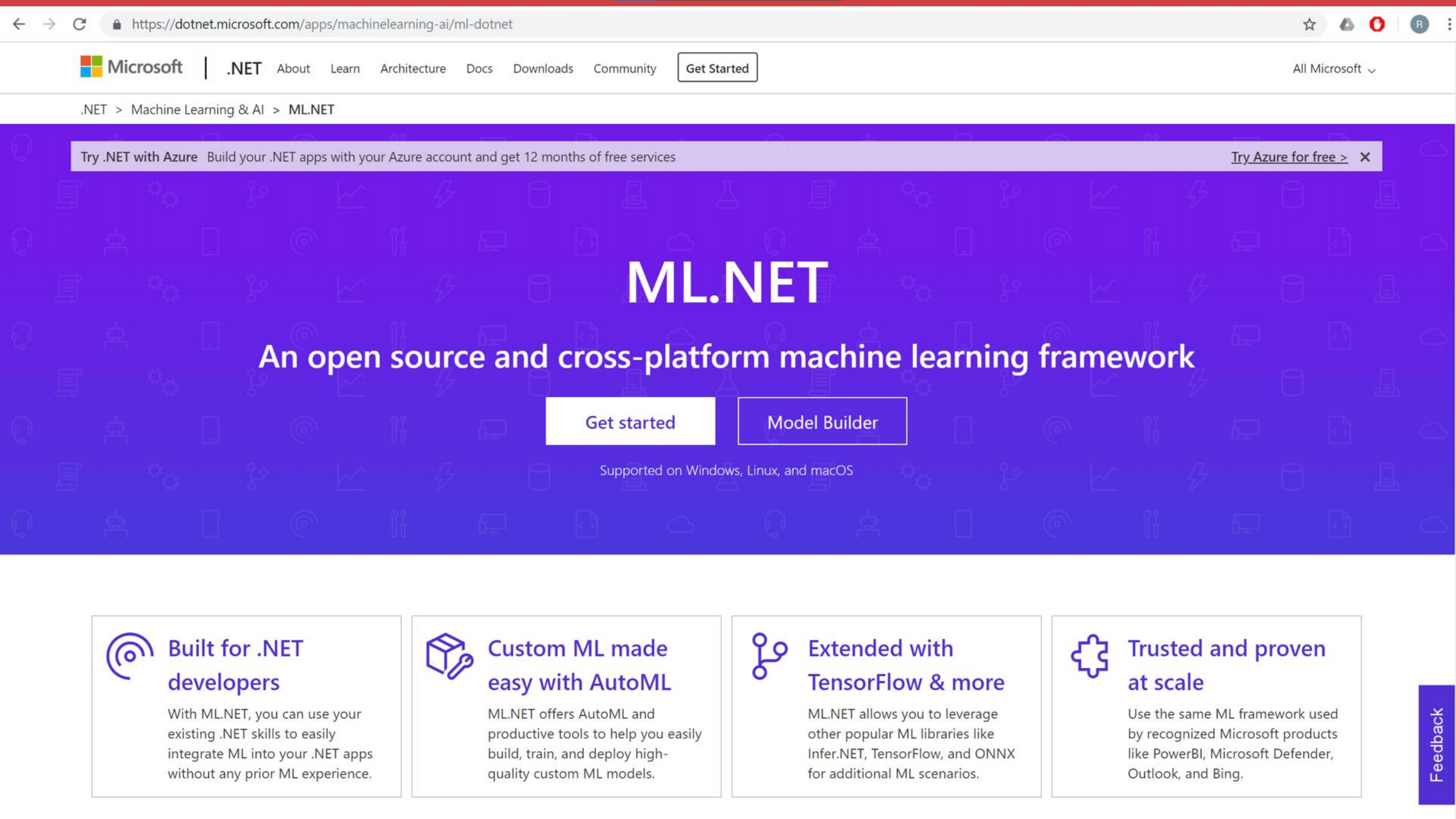


**WHY IS THERE NEVER
ENOUGH OF IT?**

FAILED DREAMS



FAILED DREAMS EVERYWHERE



.NET

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Architecture

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Get Started

All Microsoft

.NET > Machine Learning & AI > ML.NET

Try .NET with Azure Build your .NET apps with your Azure account and get 12 months of free services

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ML.NET

An open source and cross-platform machine learning framework

Get started

Model Builder

Supported on Windows, Linux, and macOS



Built for .NET developers

With ML.NET, you can use your existing .NET skills to easily integrate ML into your .NET apps without any prior ML experience.



Custom ML made easy with AutoML

ML.NET offers AutoML and productive tools to help you easily build, train, and deploy high-quality custom ML models.



Extended with TensorFlow & more

ML.NET allows you to leverage other popular ML libraries like Infer.NET, TensorFlow, and ONNX for additional ML scenarios.



Trusted and proven at scale

Use the same ML framework used by recognized Microsoft products like PowerBI, Microsoft Defender, Outlook, and Bing.

Feedback

Get started

Model

Supported on Windows, Linux, and



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Custom ML made easy with AutoML

ML.NET offers AutoML and productive tools to help you easily build, train, and deploy high-quality custom ML models.



ML.NET

- MS machine learning SDK that works offline
- Simple yet powerful
- Already powering Microsoft products
- Support major ML models
 - TensorFlow
 - ONNX
- Awesome samples on GitHub
 - <https://github.com/dotnet/machinelearning-samples>



My goal

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Education



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My story

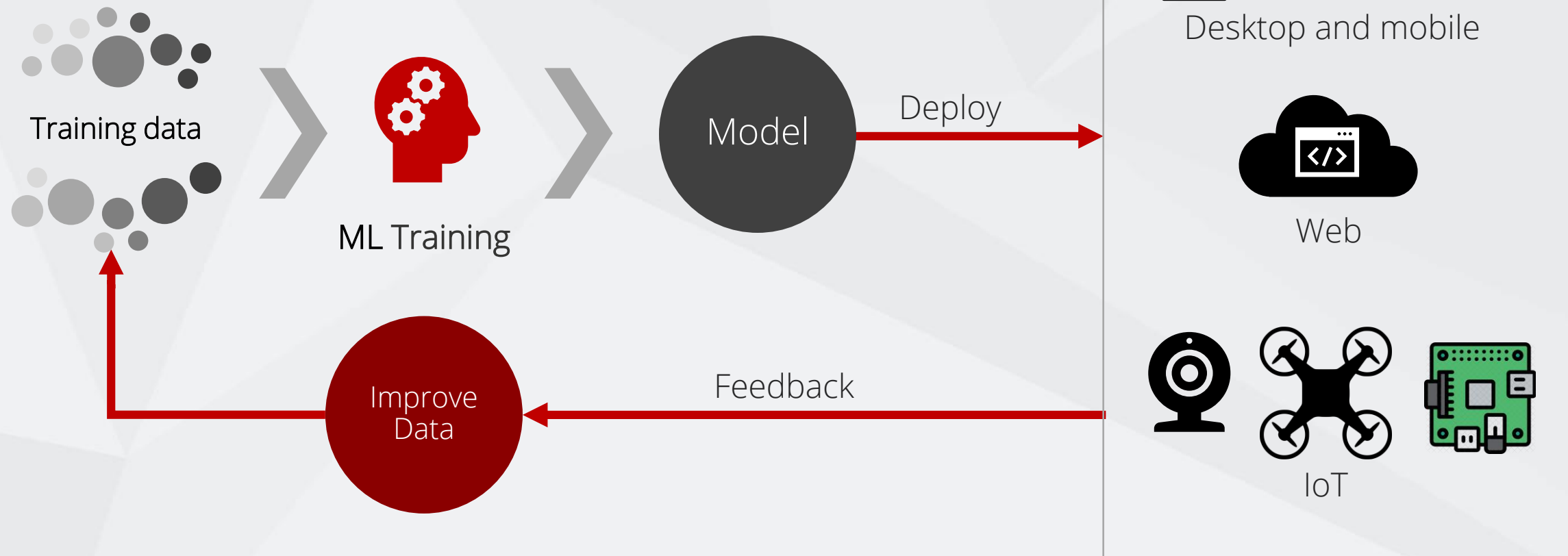
What is machine learning

ML.NET Model Builder

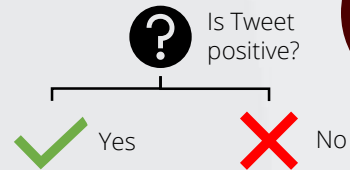
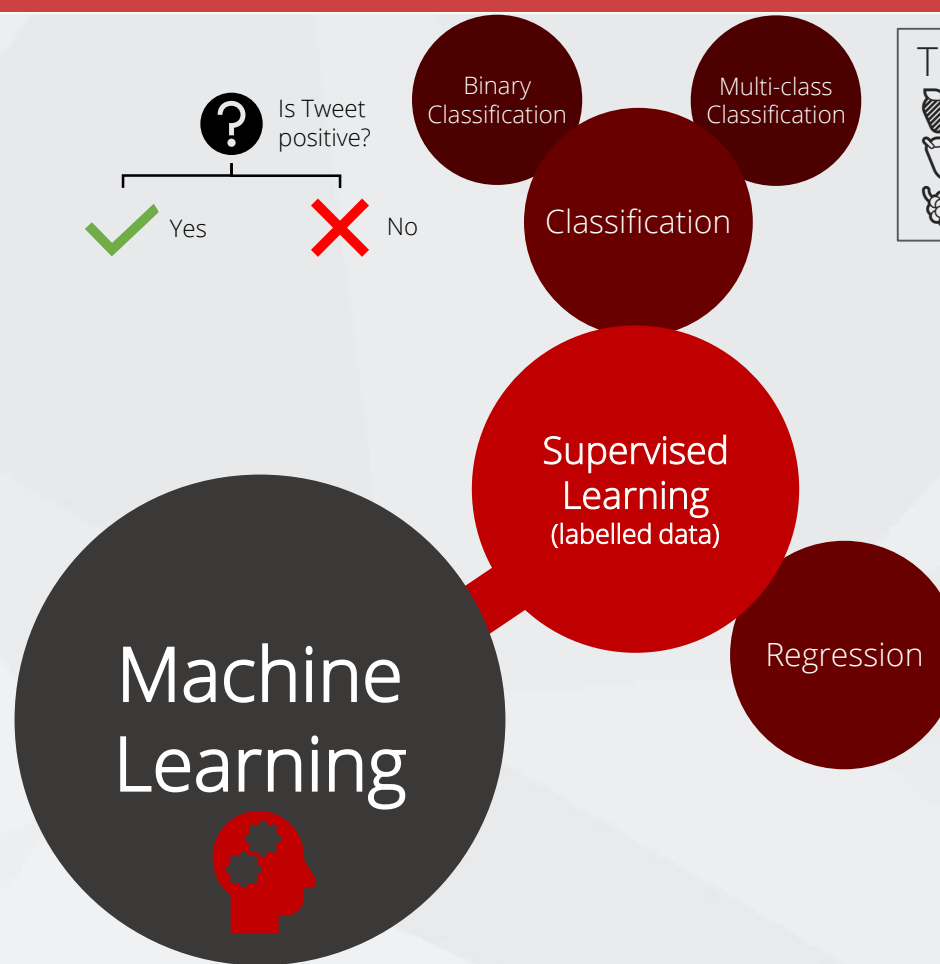
What is Machine Learning?



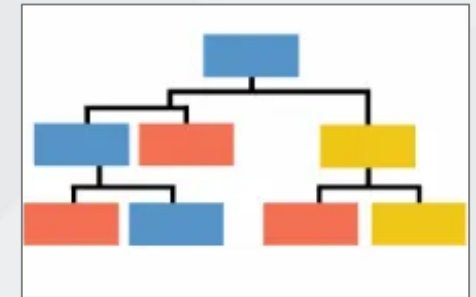
Machine Learning process



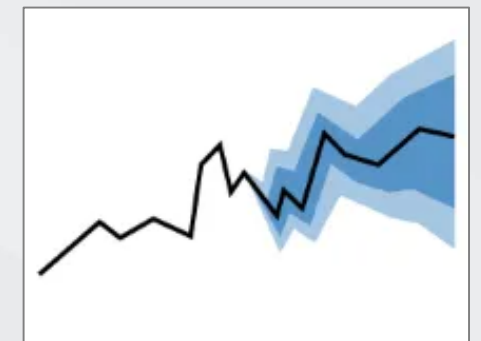
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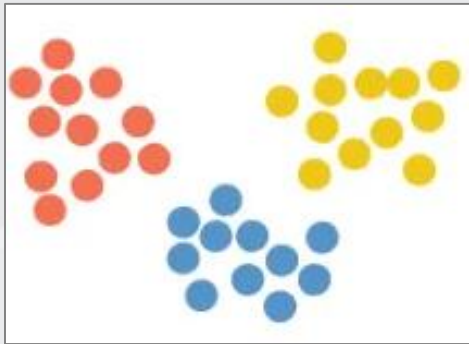
Training data:
🍏 => Apple
🍐 => Pear
🍇 => Grape



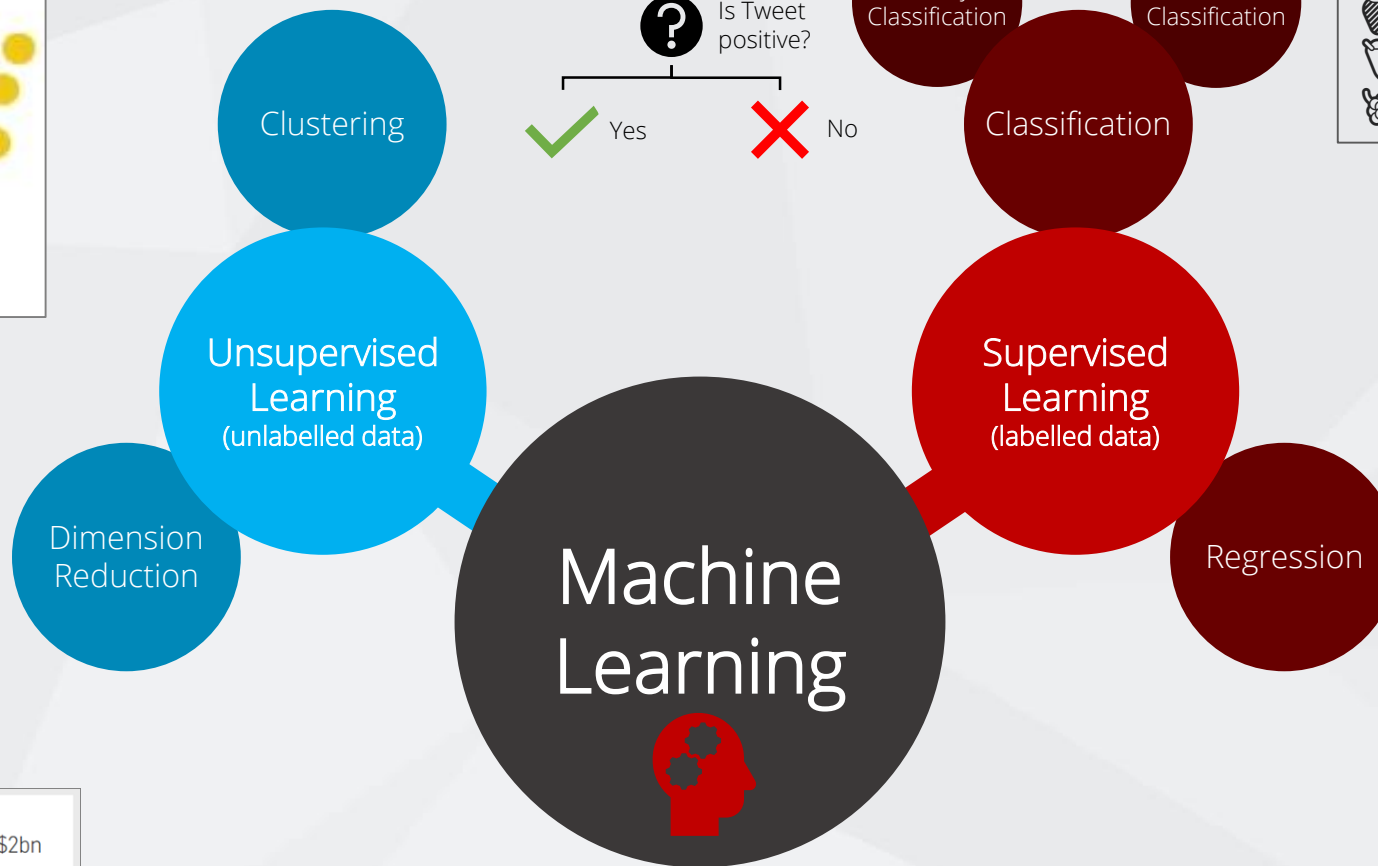
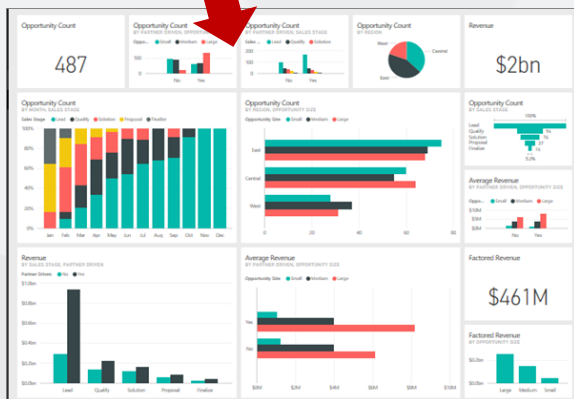
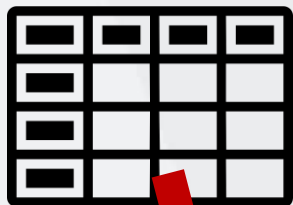
Decision trees



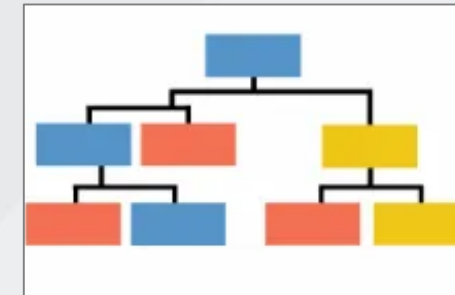
House prices predictions



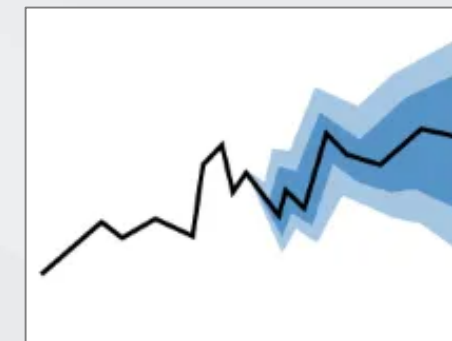
Product recommendations



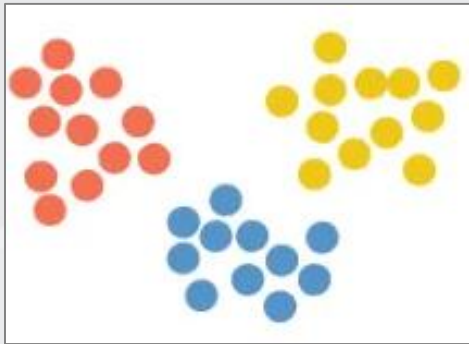
Training data:
 => Apple
 => Pear
 => Grape



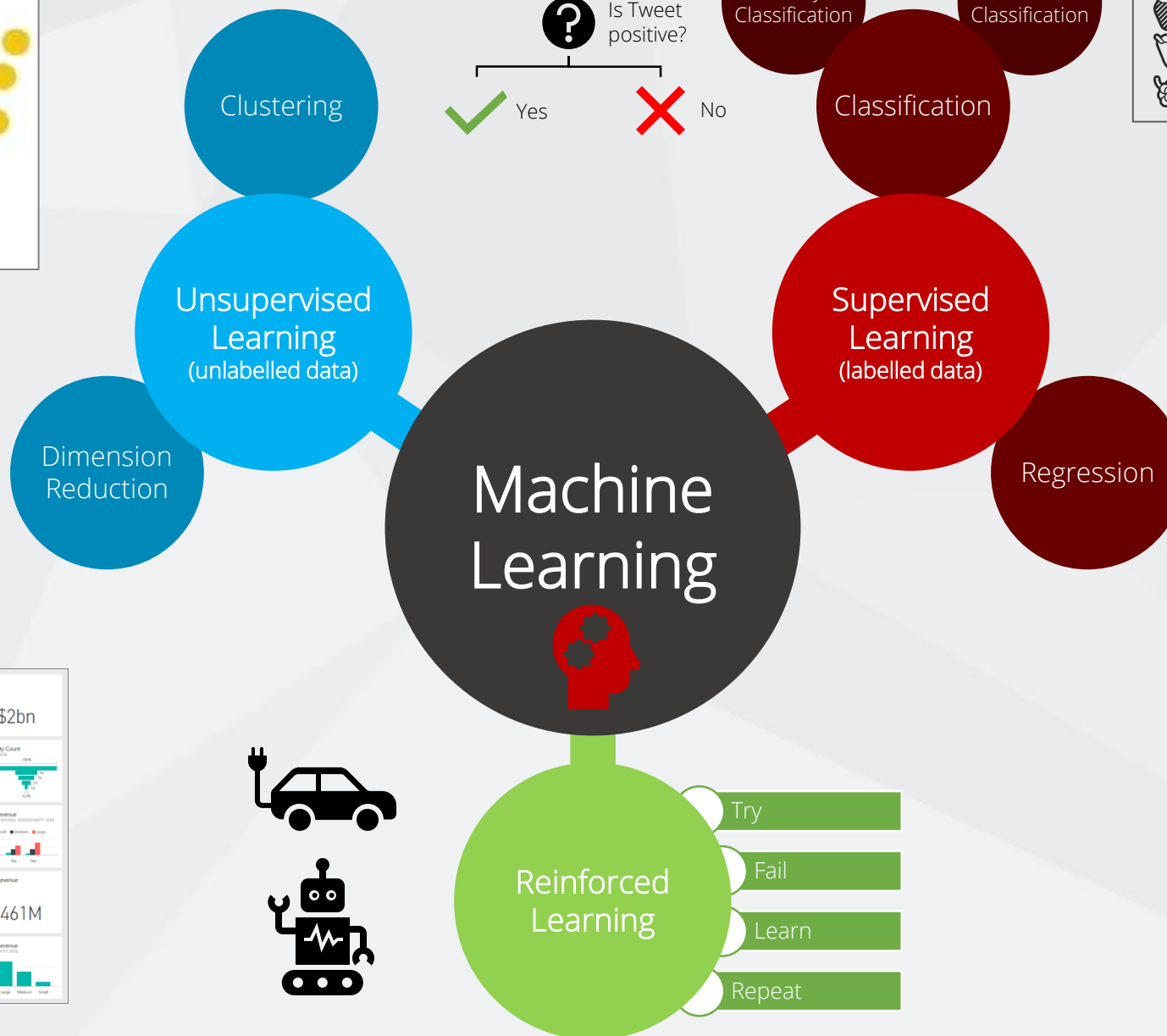
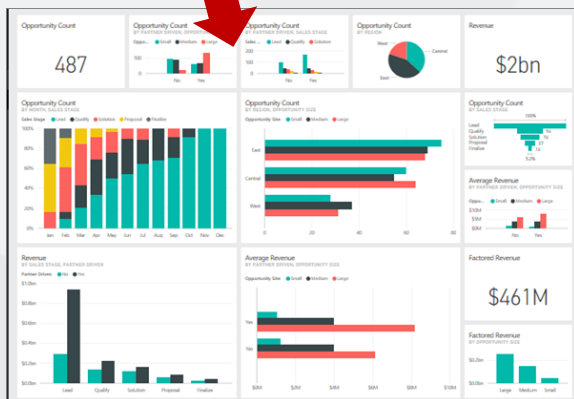
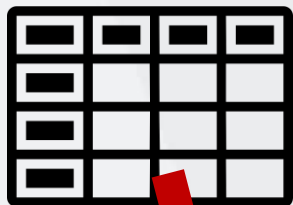
Decision trees



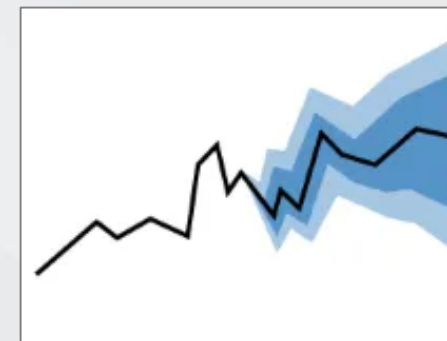
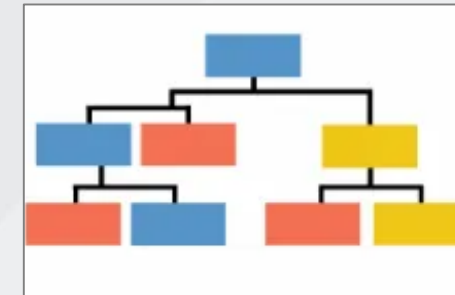
House prices predictions



Product recommendations



Training data:
 => Apple
 => Pear
 => Grape



House prices predictions

Machine Learning

Supervised Learning
(labelled data)

Classification

Multi-class Classification

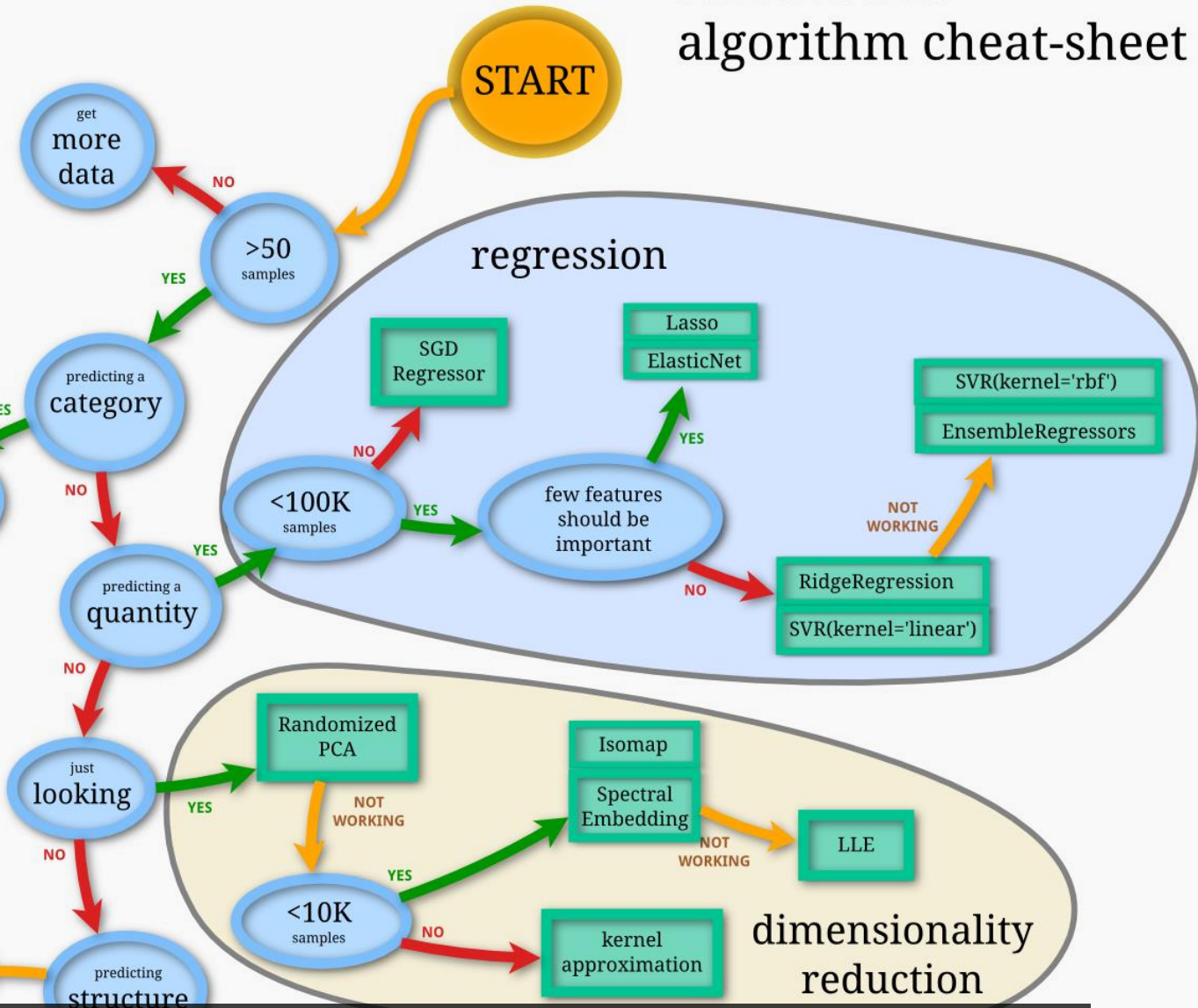
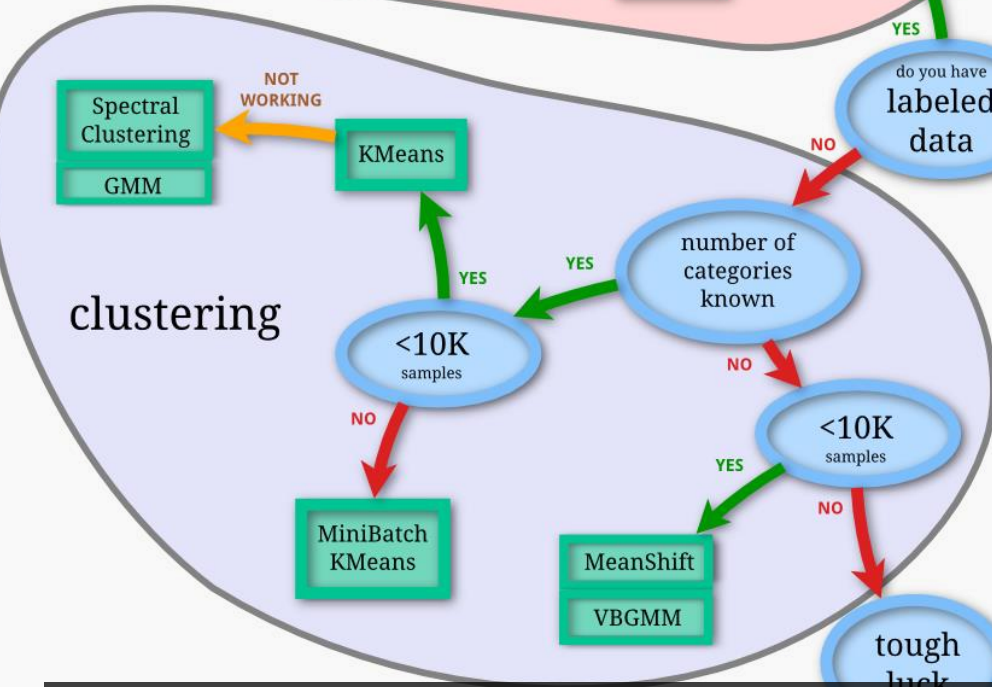
Multiclass classification trainers

You can train a multiclass classification model using the following training algorithms:

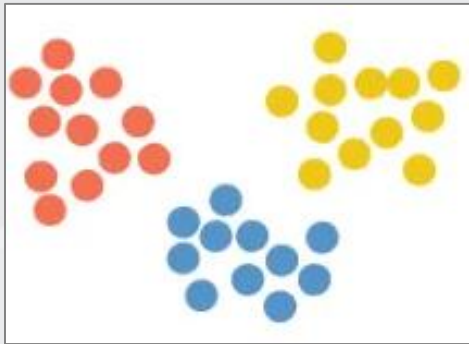
- `LightGbmMulticlassTrainer`
- `SdcaMaximumEntropyMulticlassTrainer`
- `SdcaNonCalibratedMulticlassTrainer`
- `LbfgsMaximumEntropyMulticlassTrainer`
- `NaiveBayesMulticlassTrainer`
- `OneVersusAllTrainer`
- `PairwiseCouplingTrainer`

<https://docs.microsoft.com/en-us/dotnet/machine-learning/resources/tasks>

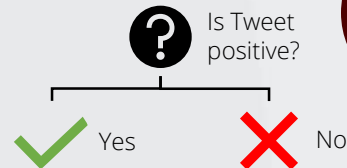
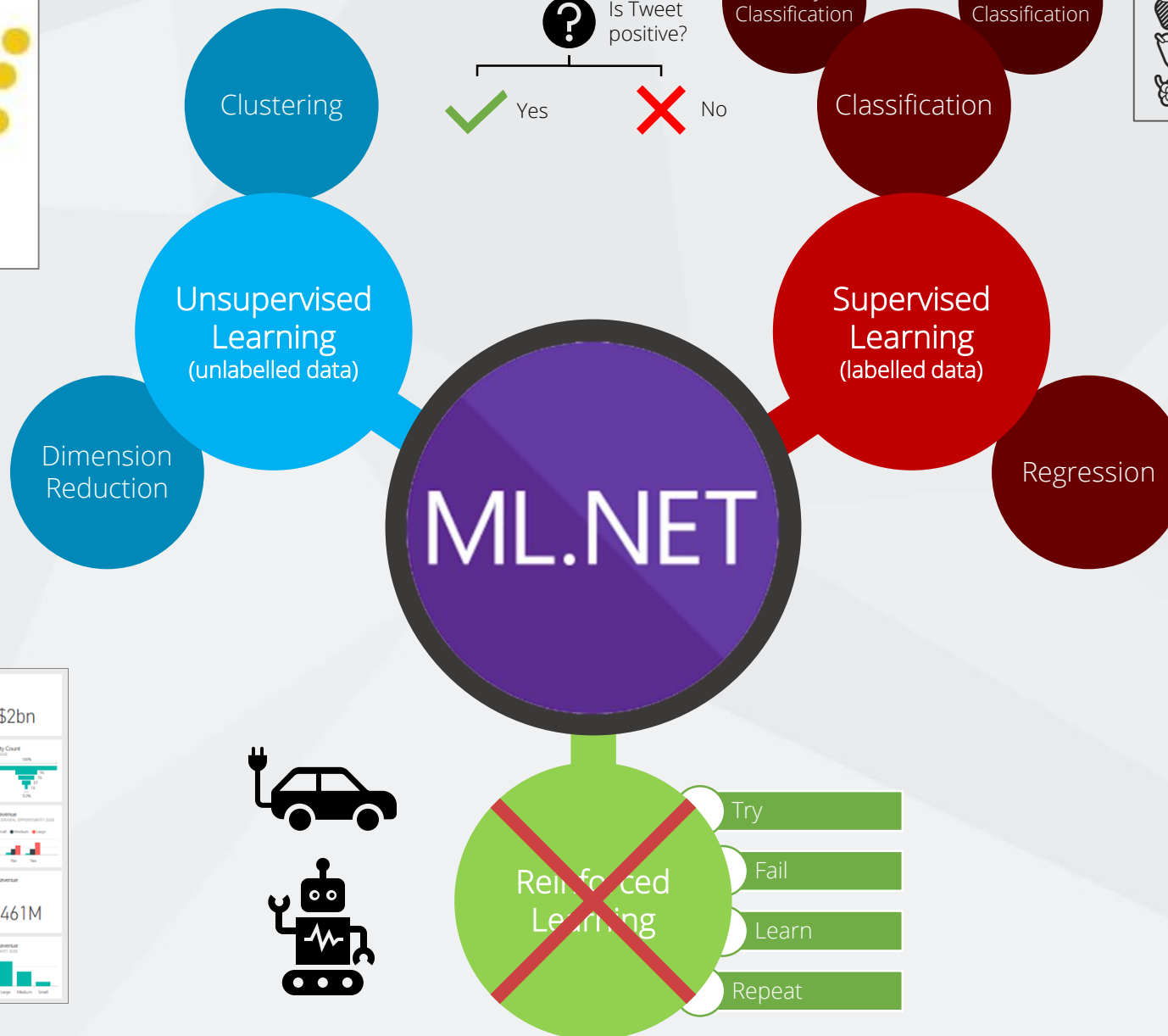
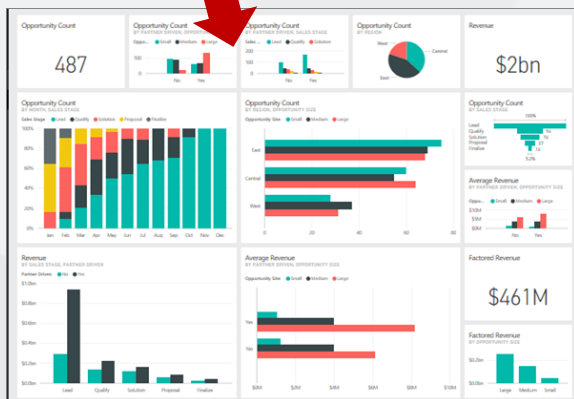
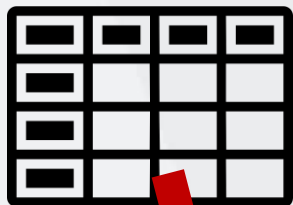
classification



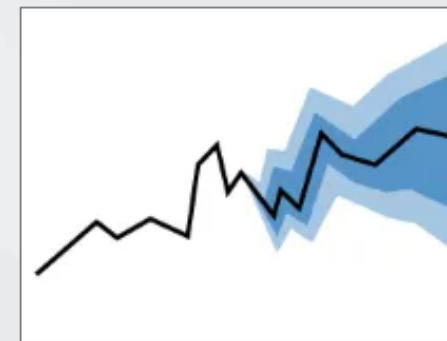
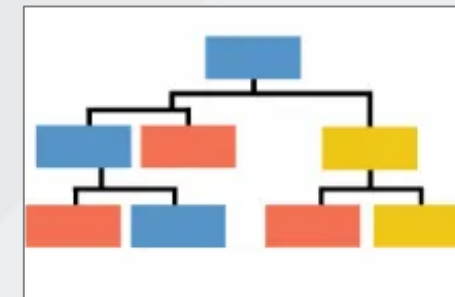
https://scikit-learn.org/stable/tutorial/machine_learning_map/index.html



Product recommendations



Training data:
 => Apple
 => Pear
 => Grape



House prices predictions

How do I start?

- Start with a scenario

Scenarios

- Is Tweet positive?
Binary classification
- Categorize bank transactions
Multi-class classification
- House price predictions
Regression
- Product recommendation
Clustering

Caveats

- ML won't magically solve your problem
 - Understand your data
 - Guide ML toward right solution
 - Start with something simple like classification



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An open source and cross-platform machine learning

Get started

Model Builder

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Extended with TensorFlow &

ML.NET allows you to leverage other popular ML libraries like Infer.NET, TensorFlow, and PyTorch for additional ML scenarios.

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ML.NET scenarios



Sentiment analysis

Analyze the sentiment of customer reviews using a binary classification algorithm.



Product recommendation

Recommend products based on purchase history using a matrix factorization algorithm.



Price prediction

Predict taxi fares based on distance traveled etc. using a regression algorithm.



Customer segmentation

Identify groups of customers with similar profiles using a clustering algorithm.



Object detection

Recognize objects in an image using an ONNX deep learning model.



Fraud detection

Detect fraudulent credit card transactions using a binary classification algorithm.



Sales spike



Image

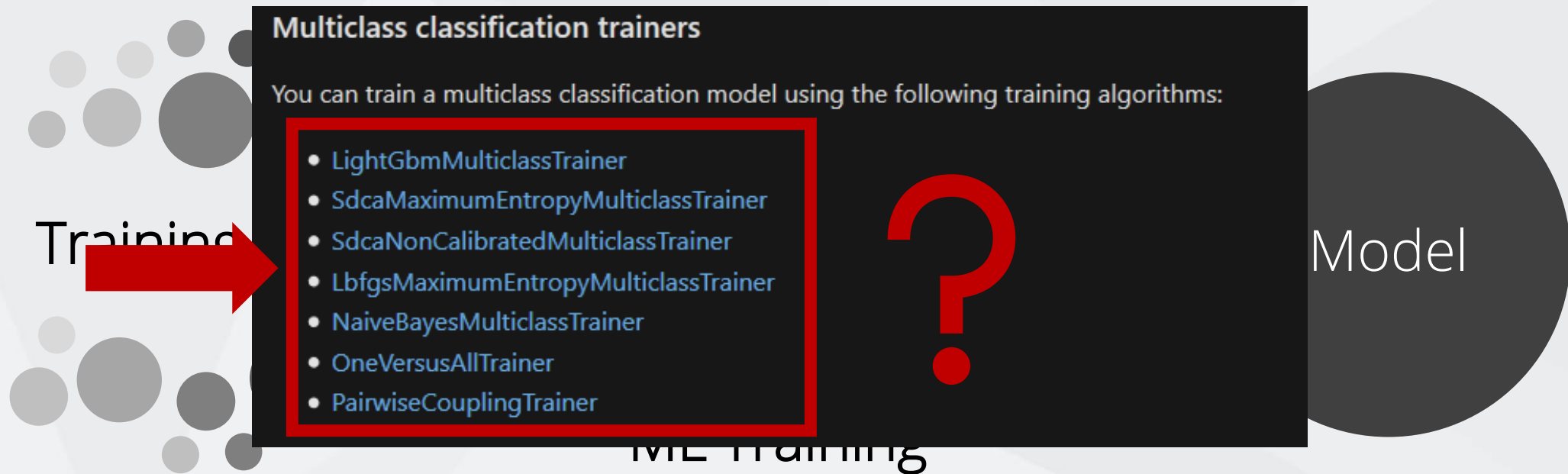


Sales forecasting

Build model (acquire knowledge)

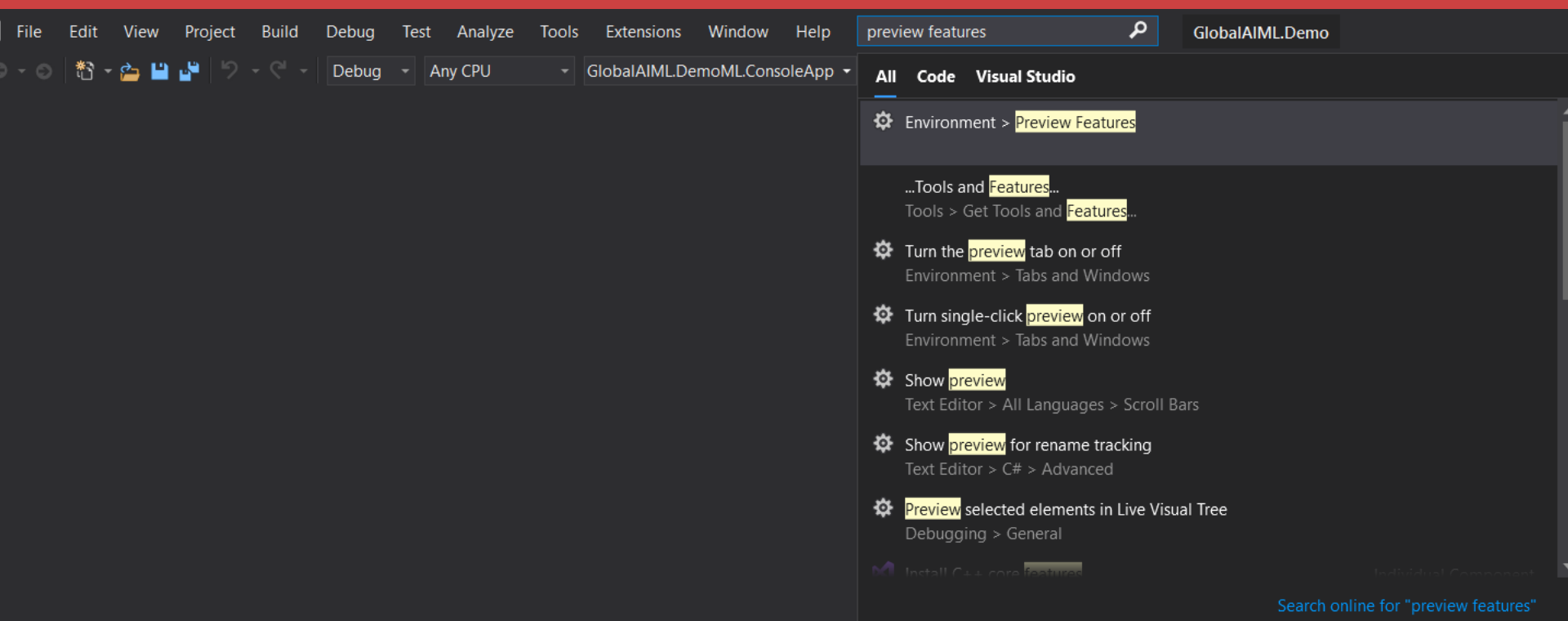


Build model (acquire knowledge)

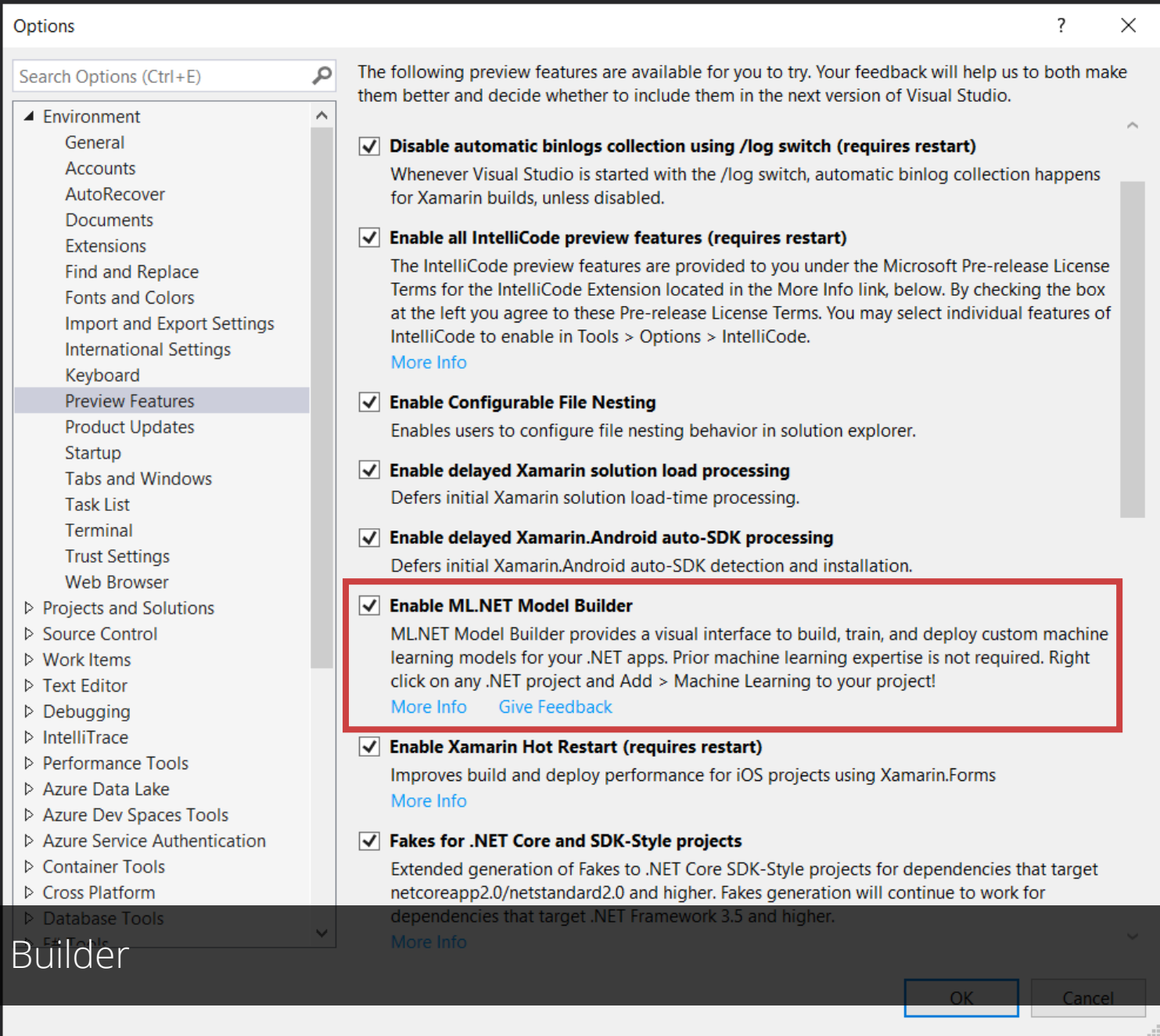


ML.NET Model Builder

- Wizard built on top of ML.NET SDK
- Picks best trainer for given scenario and data
 - Can be different one as the data changes over time
- Allows quick prototyping



Search for preview features



Enable ML.NET Model Builder

Let's do it

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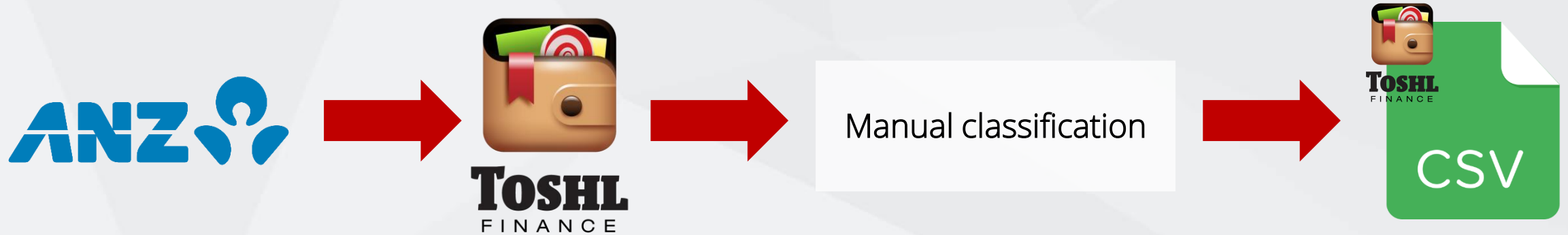


Education





















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Training data preparation



Training data (CSV)

Date	Account	Category	Tags	Expense am	Income am	Currency	In main cur	Main curre	Description						
17/05/2018	ANZ Rewards	 Leisure	movies & music	11.99	0	AUD	11.99	AUD	Spotify P05F97AC4B Sydney						
17/05/2018	Online Shopping	 Food & Drinks	 restaurants	15.15	0	AUD	15.15	AUD	VISA DEBIT PURCHASE CARD 0012 NAVALA CHURRASCARIA BRISBANE						
17/05/2018	ANZ Rewards	 Transport	 taxi	23.25	0	AUD	23.25	AUD	PAYPAL *UBER AU 4029357733						
17/05/2018	Online Shopping	 Food & Drinks	 coffee & tea	5.5	0	AUD	5.5	AUD	VISA DEBIT PURCHASE CARD 0012 AMERICAN CONCEPTS PT BRISBANE						
17/05/2018	Online Shopping	 Food & Drinks	 bar	7	0	AUD	7	AUD	VISA DEBIT PURCHASE CARD 0012 RIVERLAND BRISBANE BRISBANE CITY						
17/05/2018	Online Shopping	 Food & Drinks	 coffee & tea	24.91	0	AUD	24.91	AUD	VISA DEBIT PURCHASE CARD 0012 AT BROTHERS PTY LTD BRISBANE CITY						
17/05/2018	Online Shopping	 Food & Drinks	 groceries	19	0	AUD	19	AUD	VISA DEBIT PURCHASE CARD 0012 RAW GROUP STAFFORD P STAFFORD						
17/05/2018	Online Shopping	 Leisure	movies & music	17.25	0	AUD	17.25	AUD	VISA DEBIT PURCHASE CARD 0012 THE HOYTS CORPORATIO STAFFORD						
17/05/2018	Online Shopping	 Food & Drinks	 groceries	9.5	0	AUD	9.5	AUD	VISA DEBIT PURCHASE CARD 0012 RAW GROUP STAFFORD P STAFFORD						
17/05/2018	Online Shopping	 Food & Drinks	 restaurants	15.15	0	AUD	15.15	AUD	VISA DEBIT PURCHASE CARD 0012 NAVALA CHURRASCARIA BRISBANE						

1. Scenario

2. Environment

3. Data

4. Train

5. Evaluate

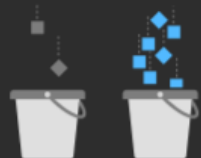
6. Code

Select a scenario

Train with your data

The following scenarios use Automated ML to train and pick the best model for your data.

[Learn more about training with your own data in Model Builder.](#)



Text classification

Classify text data into 2+ categories, e.g. predict if comments are positive or negative sentiments.

Local ML



Value prediction

Predict a numeric value from your data (regression), e.g. predict the price of a house based on features like size, location, etc.

Local ML



Image classification

Classify images into 2+ categories, e.g. predict whether an image is of a dog or a cat.

Azure ML

Local ML



Recommendation

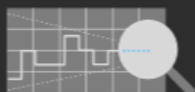
Produce a list of suggested items for a particular user, e.g. recommend products.

Local ML

Limited scenarios

The following scenarios are not yet supported by Automated ML, so walkthroughs with an example dataset and pre-defined training code are provided.

[Learn more about examples in Model Builder.](#)



Anomaly detection

Detect abnormalities or outliers in data. This example detects spikes in shampoo sales.



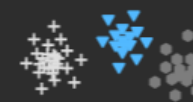
Forecasting

Predict future values based on previously observed time series values. This example predicts bike rental service demand.



Object detection

Detect and identify objects in images. This example detects objects (such as boats, people, and sofas) in images and draws



Clustering

Identify groups of related items without any pre-existing labels or categories. This example divides a set of iris flowers into different



1. Scenario



2. Environment

3. Data

4. Train

5. Evaluate

6. Code

Select training environment

This scenario only supports a local training environment.



Local

Train locally on your machine.

Local machine configuration

Processor AMD Ryzen 5 3600 6-Core Processor

Memory 33GB

Next Step: get your data

Data



1. Scenario



2. Environment

3. Data

4. Train

5. Evaluate

6. Code

Add data

In order to build a model, you must add data and choose your column to predict.
[How do I get sample datasets and learn more?](#)


Input

Choose input data source from either SQL Server or File:


File

Select a file:

Supported file formats: .csv, .tsv or .txt.

Column to predict (Label): 

Select column

Input Columns (Features): 

Select column(s)

Data PreviewData preview

Select data to see the preview.

- ✓ 1. Scenario
- ✓ 2. Environment
- ✓ 3. Data
- 4. Train
- 5. Evaluate
- 6. Code

Add data

In order to build a model, you must add data and choose your column to predict.
[How do I get sample datasets and learn more?](#)

Input

Choose input data source from either SQL Server or File:

File

Select a file: C:\DataJK\ML\from-bank-unfiltere ...

Supported file formats: .csv, .tsv or .txt.

Column to predict (Label): ⓘ

Category

Input Columns (Features): ⓘ

1 of 8 columns selected

Data PreviewData preview

10 of 2,553 rows and 2 of 9 columns. (1 Label, 1 Feature).

Category (Label)	Description
Leisure	Spotify P05F97AC4B Sydney
Other	AMAZON MKTPLC AU SYDNEY SOUTH
Transport	PAYPAL *UBER AU 4029357733
Transport	PAYPAL *UBER AU 4029357733
Other	AMAZON MKTPLC AU SYDNEY SOUTH
Food & Drinks	VISA DEBIT PURCHASE CARD 0012 AT BROTHERS PTY LTD BRISBANE CITY
Food & Drinks	VISA DEBIT PURCHASE CARD 0012 RAW GROUP STAFFORD P STAFFORD
Leisure	VISA DEBIT PURCHASE CARD 0012 THE HOYTS CORPORATIO STAFFORD
Food & Drinks	VISA DEBIT PURCHASE CARD 0012 RAW GROUP STAFFORD P STAFFORD
Food & Drinks	VISA DEBIT PURCHASE CARD 0012 NAVALA CHURRASCARIA BRISBANE

Next Step: train your model

Train



1. Scenario



2. Environment



3. Data

4. Train

5. Evaluate


6. Code

Train

Specify a time to train for evaluating various models.

[How long should I train for?](#)

Input

Time to train (seconds): 

10

Start training

Progress

Start training to see progress and results.

- ✓ 1. Scenario
- ✓ 2. Environment
- ✓ 3. Data

4. Train

5. Evaluate

6. Code

Train

Specify a time to train for evaluating various models.
[How long should I train for?](#)

Input

Time to train (seconds): ⓘ

10

Cancel training

Progress

Status:



2 seconds remaining...

Best accuracy:

84.71%

Best algorithm:

AveragedPerceptronOva

- ✓ 1. Scenario
- ✓ 2. Environment
- ✓ 3. Data
- ✓ 4. Train
- 5. Evaluate
- 6. Code

Train

Specify a time to train for evaluating various models.
[How long should I train for?](#)

Input

Time to train (seconds): ⓘ

Start training

Progress

Status: ✓ Training complete
Best accuracy: 84.53%
Best algorithm: AveragedPerceptronOva

Next step: evaluate your model

Evaluate

Output

Show output from: Machine Learning



Total experiment time : 49.5641089 Secs					
Total number of models explored: 6					

Top 5 models explored					

	Trainer	MicroAccuracy	MacroAccuracy	Duration	#Iteration
1	AveragedPerceptronOva	0.8453	0.7189	6.1	1
2	SdcaMaximumEntropyMulti	0.8430	0.6074	5.0	2
3	LinearSvmOva	0.8264	0.6760	3.6	3
4	SymbolicSgdLogisticRegressionOva	0.8175	0.5084	6.3	4
5	LightGbmMulti	0.6615	0.3155	5.3	5

- ✓ 1. Scenario
- ✓ 2. Environment
- ✓ 3. Data

✓ 4. Train

5. Evaluate

6. Code

Train

Specify a time to train for evaluating various models.
[How long should I train for?](#)

Input

Time to train (seconds): ⓘ

60

Start training

Progress

Status: ✓ Training complete
Best accuracy: 84.53%
Best algorithm: AveragedPerceptronOva

Next step: evaluate your model

Evaluate

Output

Show output from: Machine Learning

| Total experiment time : 49.5641089 Secs
| Total number of models explored: 6

Top 5 models explored

	Trainer	MicroAccuracy	MacroAccuracy	Duration	#Iteration
1	AveragedPerceptronOva	0.8453	0.7189	6.1	1
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3	LinearSvmOva	0.8264	0.6760	3.6	3
4	SymbolicSgdLogisticRegressionOva	0.8175	0.5084	6.3	4
5	LightGbmMulti	0.6615	0.3155	5.3	5

- ✓ 1. Scenario
- ✓ 2. Environment
- ✓ 3. Data
- ✓ 4. Train
- ✓ 5. Evaluate
- 6. Code

Evaluate

Results of training for your model can be found below. [How do I understand my model performance?](#)

Output

Overview Details

ML task: multiclass-classification
Training time: 49.56 seconds
Models explored (total): 6 | [View Top 5 models explored](#)

Overall accuracy:

84.53%

Try your model

Sample data

Note: Fields below are pre-filled by row 1 of your data.

Description






Spotify P05F97AC4B Sydney

Predict

Next step: Get your code snippet

Code

Top 5 Results:

 Leisure	95%
 Food & Drinks	2%
 Home & Utilities	1%
 Other	1%
 Transport	1%

- ✓ 1. Scenario
- ✓ 2. Environment
- ✓ 3. Data
- ✓ 4. Train
- ✓ 5. Evaluate

6. Code

Code

Add the machine learning model and the projects and references for model consumption (**GlobalAIML.DemoML.Model**) and training/testing (**GlobalAIML.DemoML.ConsoleApp**) to your solution.

Add Projects

✓ 1. Scenario

✓ 2. Data

✓ 3. Train

✓ 4. Evaluate

✓ 5. Code

Code

Add the machine learning model and the projects and references for model consumption (**GlobalAIML.DemoML.Model**) and training/testing (**GlobalAIML.DemoML.ConsoleApp**).

Added Projects

Next Steps

1. Try the model

Run **GlobalAIML.DemoML.ConsoleApp** to make predictions on sample data.

2. Consume the model

In **GlobalAIML.Demo**, add the following using directive in the file where you will consume your model:

```
using GlobalAIML_DemoML.Model;
```

Then add your input data and use `ConsumeModel.Predict()` to load the model and predict the output:

```
// Add input data
var input = new ModelInput();

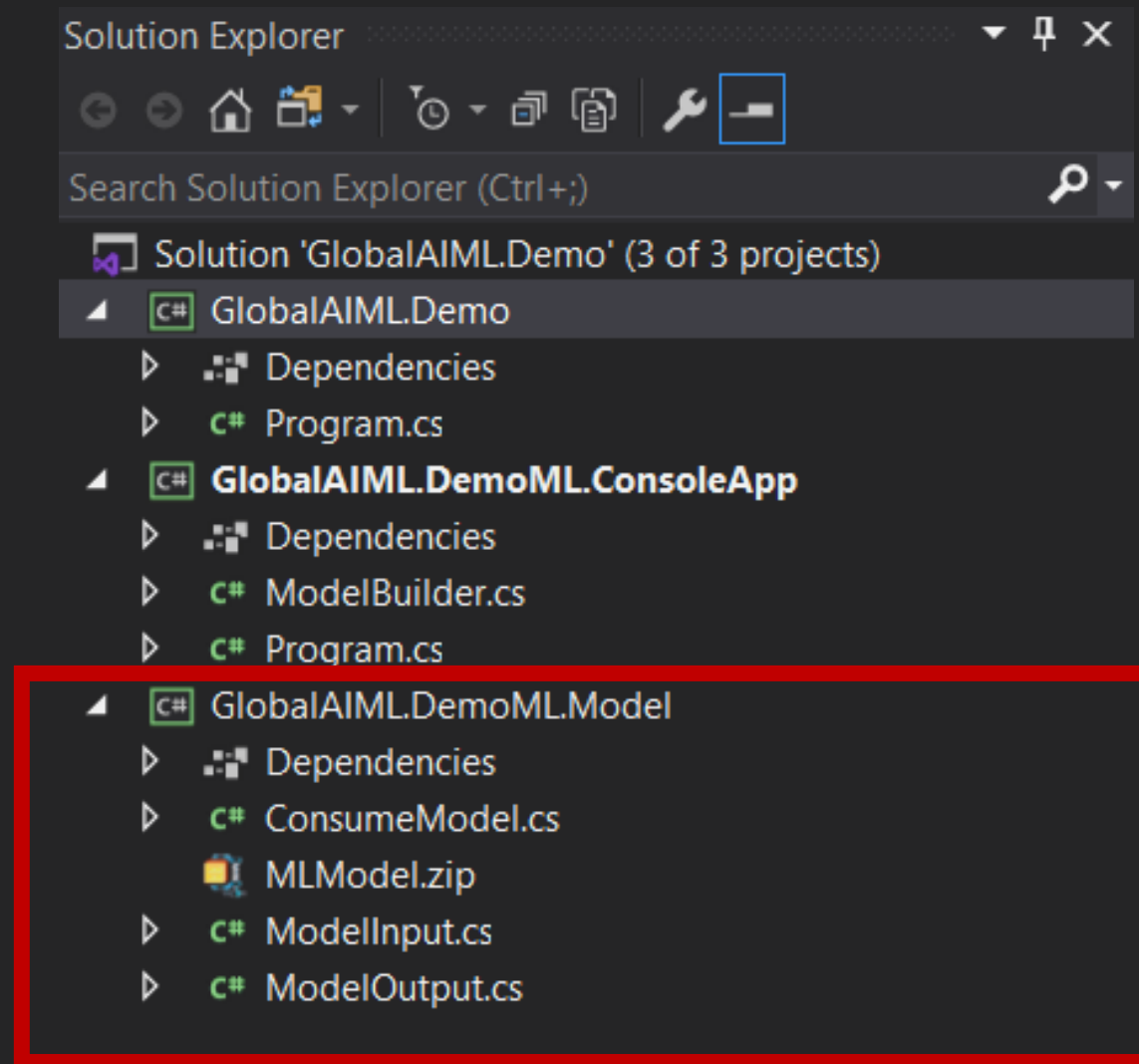
// Load model and predict output of sample data
ModelOutput result = ConsumeModel.Predict(input);
```

Check out [this tutorial](#) to learn more about how to use the model in your app.

3. Improve the model

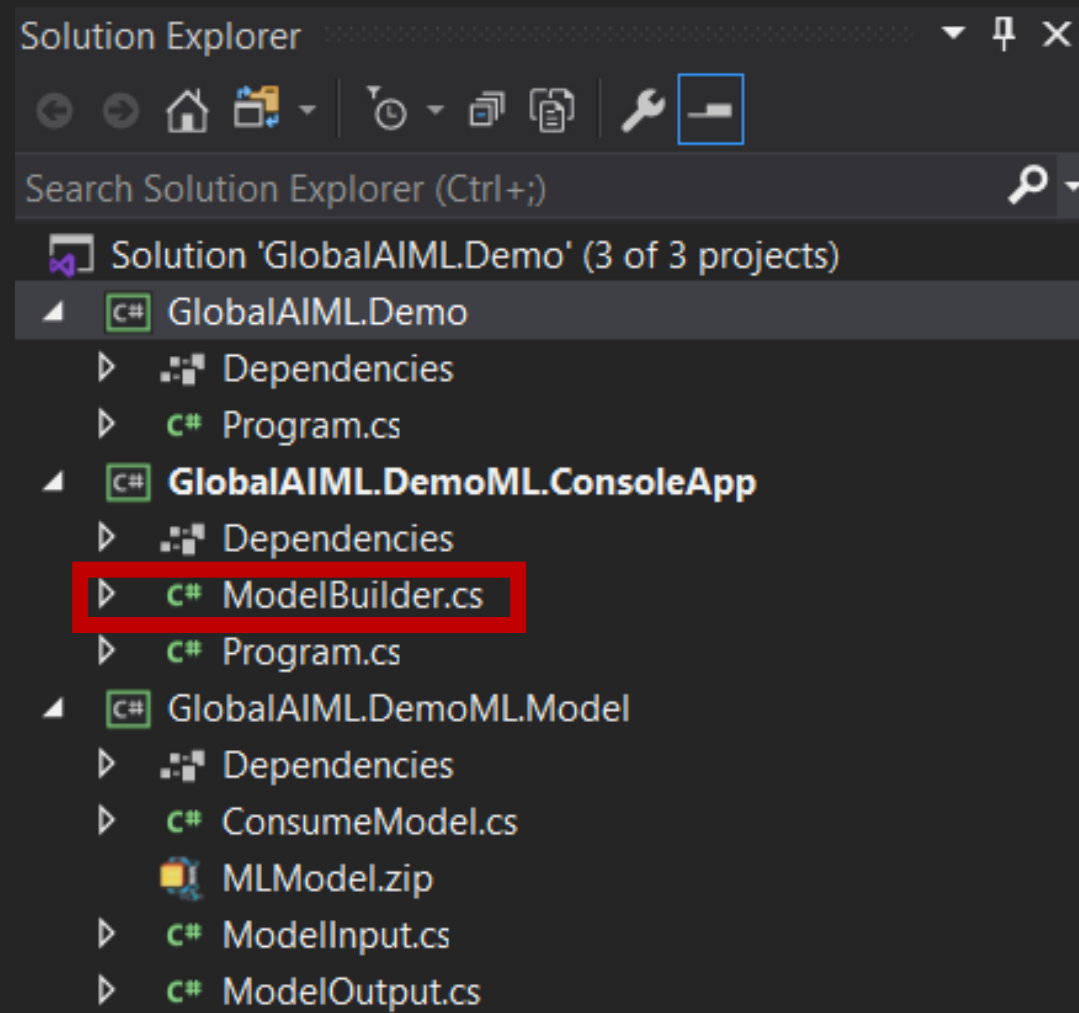
Train for a longer time, add more data, or [learn more on the web](#).

Model





ML Training






Demo

Summary

- Import CSV file
- Tweak columns to get better results
- Generate project with ML
- Tweaking generated code
- Minor issues with small datasets (100+ rows)

A video frame showing two men in a studio setting. The man on the left has a beard and glasses, wearing a blue button-down shirt. The man on the right has glasses and is wearing a black t-shirt. They are both smiling and looking at each other. In the background, a screen displays 'NDC { Sydney }'.

NDC { Sydney }

Machine Learning that's **ACTUALLY** easy

More details - <http://bit.ly/jk-ml-interview-blog>



VIDEOS BY DEVELOPERS, FOR DEVELOPERS

Goal achieved

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Join the Conversation #virtualmlnet @jernej_kavka

Thank you!

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