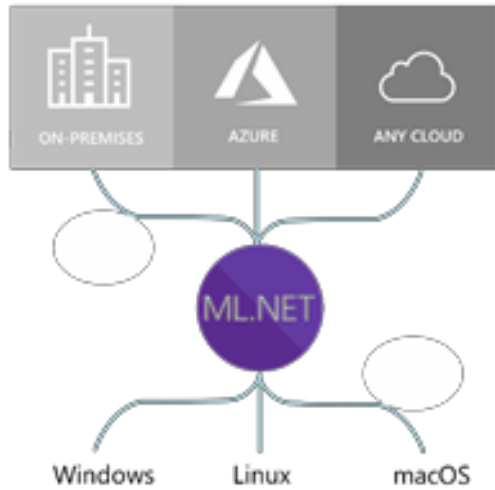


ML.NET

Conf.dr. Cristian Kevorchian
University of Bucharest

ML.NET runs anywhere



Supported Frameworks:

.NET Core (*Natively*)

.NET Framework (*Natively*)

Python with *NimbusML* (*Python bindings*)

Supported processor arch.

x64

x86

ML.NET is enterprise-ready battle proven

Microsoft Products

(Using ML.NET internally for > 8 years)



Bing - Ad Predictions



Excel - Chart Recommendations



Power Point - Design Ideas



Microsoft Defender – Antivirus Threat Protection



Azure Stream Analytics - Anomaly Detection

Customers

(ML.NET v1 since May 2019)



Evolution Software – Hazelnut drying time prediction



Williams Mullen – **Law document classification**



Sig Parser – **E-mail spam detection**



Brenmor – Medical patient survey classification



endjin – Newsletter article classification

+ more

Machine Learning complexity?

Supervised ML (Infers label)

Linear Discriminant Analysis

Structured prediction **Regression**

Naïve Bayes

Linear regression

Decision Trees

Logistic regression

Binary Classification

Multi-class Classification

k-nearest neighbor

Neural Networks

(MultiLayer Perception, etc.)

Support Vector Machines

Unsupervised ML (Infers structure)

Clustering

(K-means
Mixture models
Hierarchical clustering)

Topic Modeling

Dimensionality Reduction

Spike detection

Latent variable models

Topic modeling

Neural Networks

(Autoencoders,
Self-organizing maps, etc.)

A few things you can do with ML.NET ...



Sentiment analysis

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

[Sentiment analysis sample >](#)



Product recommendation

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

[Product recommendation sample >](#)



Price prediction

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

[Price prediction sample >](#)



Customer segmentation

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

[Customer segmentation sample >](#)



GitHub labeler

Suggest the GitHub label for new issues using a multi-class classification algorithm.

[GitHub labeler sample >](#)



Fraud detection

Detect fraudulent credit card transactions using a binary classification algorithm.

[Fraud detection sample >](#)



Spam detection

Flag text messages as spam using a binary classification algorithm.

[Spam detection sample >](#)



Image classification

Classify images (e.g. broccoli vs pizza) using a TensorFlow deep learning algorithm.

[Image classification sample >](#)



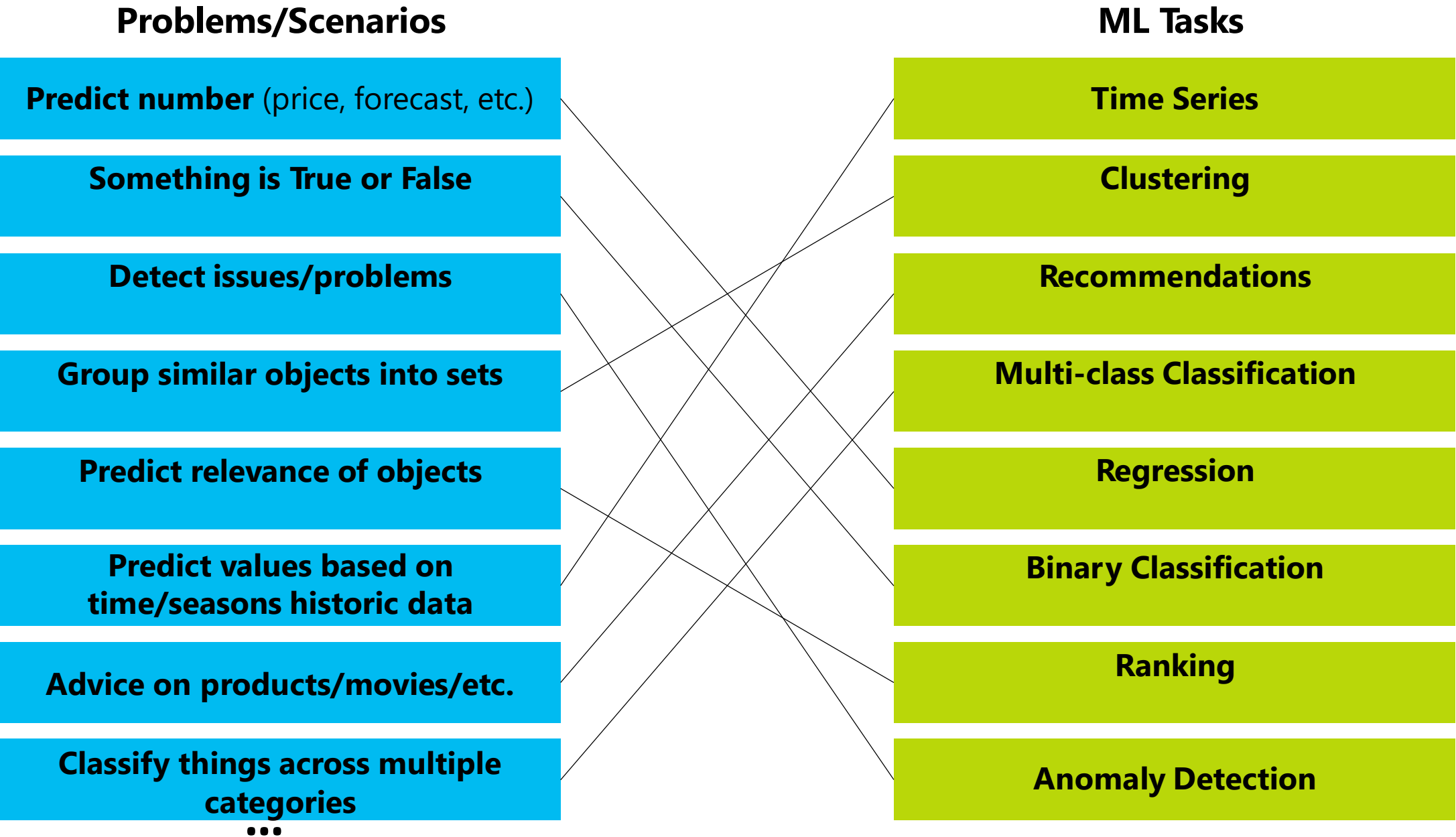
Sales forecasting

Forecast future sales for products using a regression algorithm.

[Sales forecasting sample >](#)

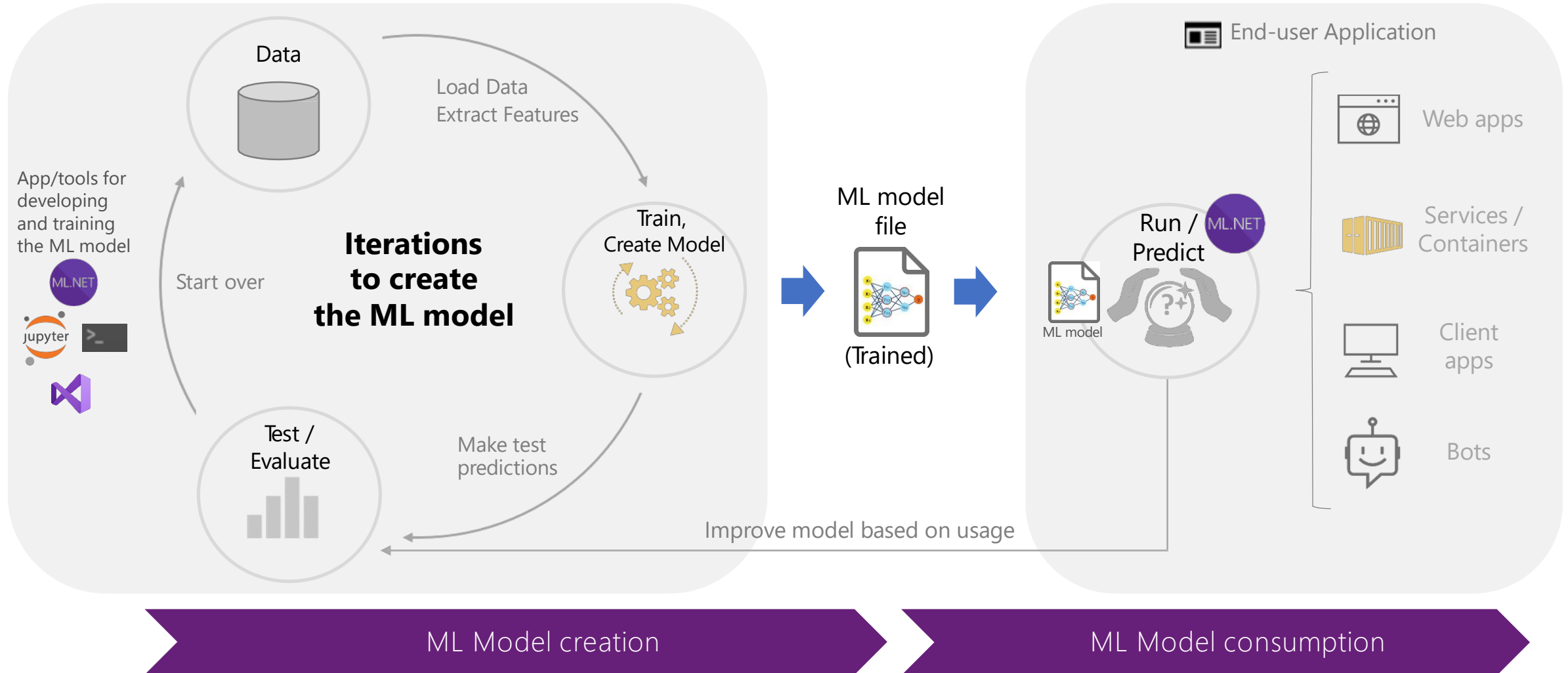
You can find more [ML.NET samples on GitHub](#), or take a look at the [ML.NET tutorials](#).

Mapping from Scenarios/Problems to ML Tasks



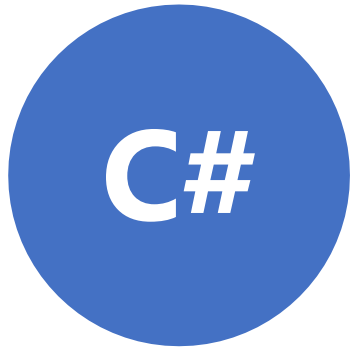
The Machine Learning journey

ML model lifecycle



Three ways to use ML.NET...

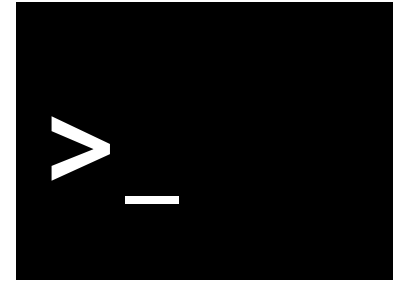
...for **training** an ML model



ML.NET
API
(Code)



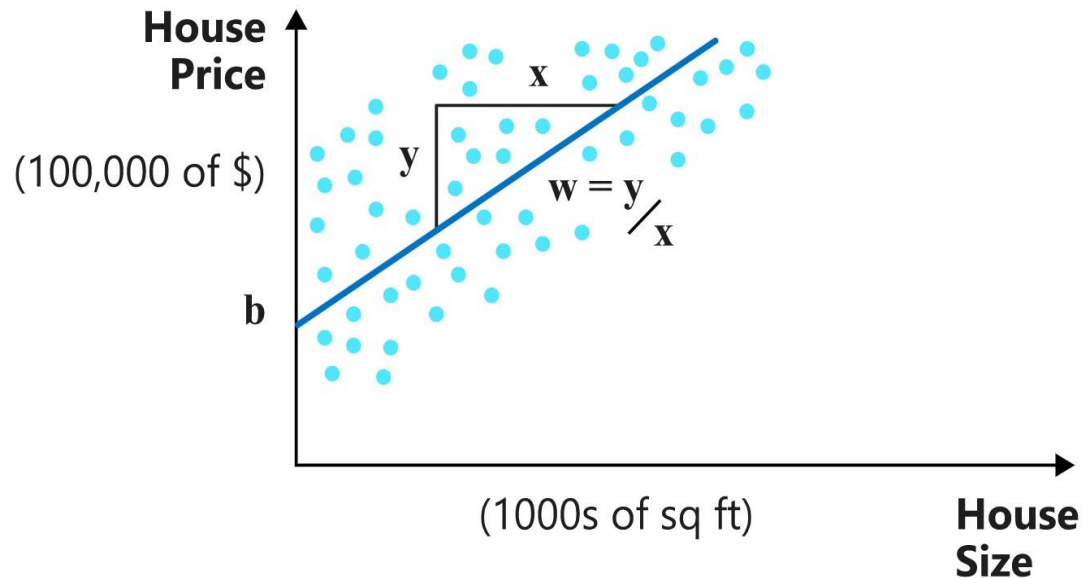
ML.NET
Model Builder
(Visual Studio UI)



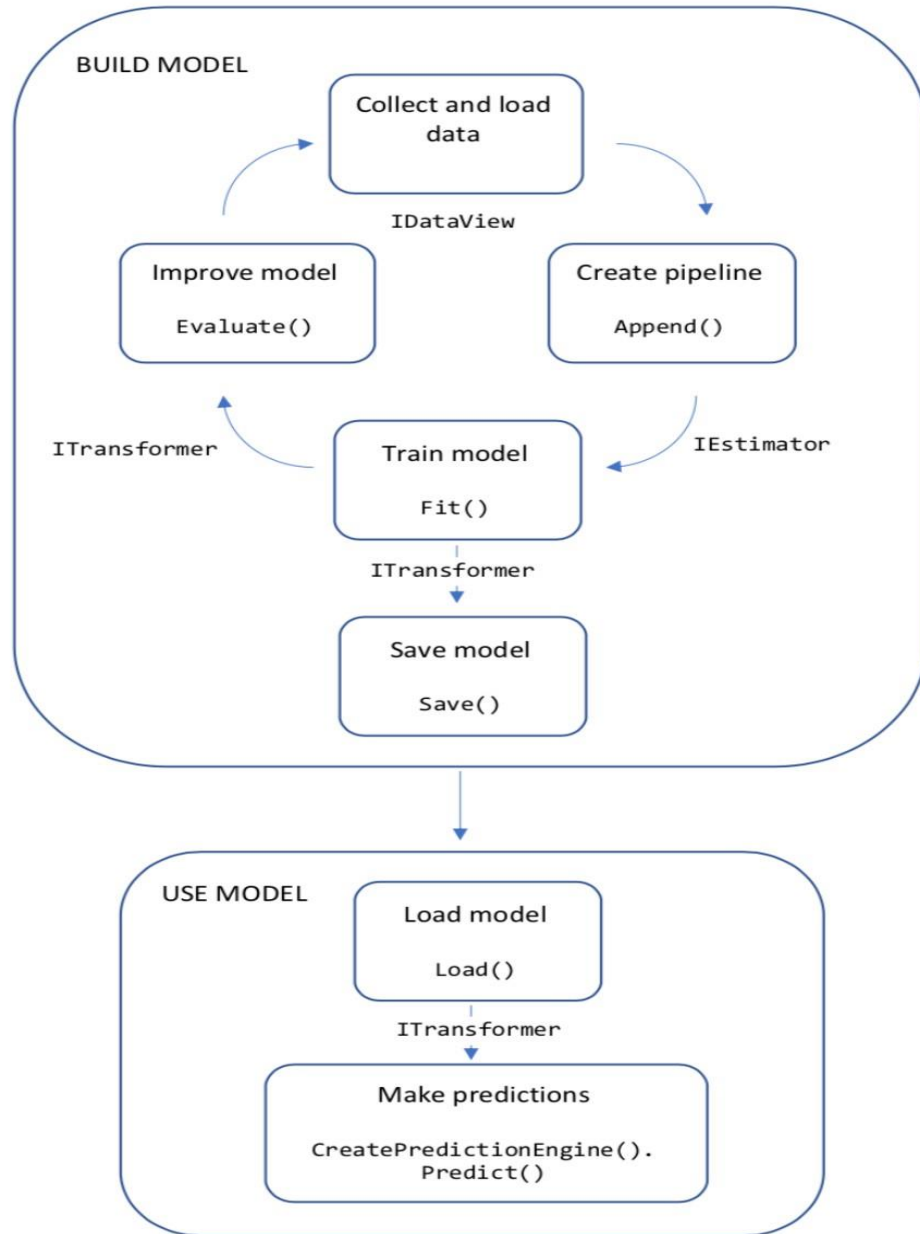
ML.NET
CLI
(Command-Line
Interface)

Recap: Machine Learning Models

- ML.NET model is an object that contains transformations to perform on your input data to arrive at the predicted output



- Most basic model is two-dimensional linear regression
- **Label** is the output, in this case Price
- **Features** are the input to the model that we will process, in this case there is only one, Size



//Step 1. Create a ML Context

```
var ctx = new MLContext();
```

//Step 2. Read in the input data for model training

```
IDataView dataReader = ctx.Data
    .LoadFromTextFile<MyInput>(dataPath, hasHeader: true);
```

//Step 3. Build your estimator

```
IEstimator<ITransformer> est = ctx.Transforms.Text
    .FeaturizeText("Features", nameof(SentimentIssue.Text))
    .Append(ctx.BinaryClassification.Trainers
        .LbfgsLogisticRegression("Label", "Features"));
```

//Step 4. Train your Model

```
ITransformer trainedModel = est.Fit(dataReader);
```

//Step 5. Make predictions using your model

```
var predictionEngine = ctx.Model
    .CreatePredictionEngine<MyInput, MyOutput>(trainedModel);
```

```
var sampleStatement = new MyInput { Text = "This is a horrible movie" };
```

```
var prediction = predictionEngine.Predict(sampleStatement);
```

Example - Analyse Sentiment

```
yelp_labelled.txt - Program.cs
1 [redacted]
2 Crust is not good. 0
3 Not tasty and the texture was just nasty. 0
4 Stopped by during the late May bank holiday off Rick Steve recommendation and loved it. 1
5 The selection on the menu was great and so were the prices. 1
6 Now I am getting angry and I want my damn pho. 0
7 Honeslty it didn't taste THAT fresh.) 0
8 The potatoes were like rubber and you could tell they had been made up ahead of time being kept i
9 The fries were great too. 1
10 A great touch. 1
11 Service was very prompt. 1
12 Would not go back. 0
13 The cashier had no care what so ever on what I had to say it still ended up being wayyy overprice
14 I tried the Cape Cod ravioli, chicken,with cranberry...mmmm! 1
15 I was disgusted because I was pretty sure that was human hair. 0
16 I was shocked because no signs indicate cash only. 0
17 Highly recommended. 1
18 Waitress was a little slow in service. 0
19 This place is not worth your time, let alone Vegas. 0
20 did not like at all. 0
```

Comment Text	Sentiment
Wow... Loved this place.	1
Crust is not good.	0
Not tasty and the texture was just nasty.	0
The selection on the menu was great.	1

Features (input)

Label (output)

```
IDataView dataView = ctx.Data.LoadFromTextFile<SentimentData>(_dataPath, hasHeader: false);
TrainTestData splitDataView = ctx.Data.TrainTestSplit(dataView, testFraction: 0.2); // create a split set
```

```
1 reference
public class SentimentData
{
    [LoadColumn(0)]
    public string SentimentText;

    [LoadColumn(1), ColumnName("Label")]
    public bool Sentiment;
}

0 references
public class SentimentPrediction : SentimentData
{
    [ColumnName("PredictedLabel")]
    0 references
    public bool Prediction { get; set; }

    0 references
    public float Probability { get; set; }

    0 references
    public float Score { get; set; }
}
```

The [redacted] method in the previous code converts the text column (SentimentText) into a numeric key type Features column used by the machine learning algorithm and adds it as a new dataset column:

SentimentText	Sentiment	Features
Waitress was a little slow in service.	0	[0.76, 0.65, 0.44, ...]
Crust is not good.	0	[0.98, 0.43, 0.54, ...]
Wow... Loved this place.	1	[0.35, 0.73, 0.46, ...]
Service was very prompt.	1	[0.39, 0, 0.75, ...]

```
IEstimator<ITransformer> est = ctx.Transforms.Text[redacted] outputColumnName: "Features", inputColumnName: nameof(SentimentData.SentimentText))
.Append(ctx.BinaryClassification.Trainers.StochasticRegression(labelColumnName: "Label", featureColumnName: "Features"));
```


Step #1 use Azure Text Analytics to extract key phrases, named entities, etc from movie scripts

```
Processing: 1_Serendipity.txt|Script file found|Chunks:10|.....|828 KP!  
Processing: 1_Sleepless in Seattle.txt|Script file found|Chunks:6|.....|511 KP!  
Processing: 1_The Holiday.txt|Script file found|Chunks:12|.....|750 KP!  
Processing: 1_When Harry Met Sally.txt|Script file found|Chunks:11|.....|717 KP!  
Processing: 2_Blair Witch Project.txt|Script file found|Chunks:7|.....|447 KP!  
Processing: 2_Saw.txt|Script file found|Chunks:5|.....|356 KP!  
Processing: 2_The Evil Dead.txt|Script file found|Chunks:2|..|152 KP!  
Processing: 2_TheShining.txt|Script file found|Chunks:8|.....|584 KP!  
Processing: 3_Baby Driver.txt|Script file found|Chunks:9|.....|757 KP!  
Processing: 3_Point Break.txt|Script file found|Chunks:8|.....|775 KP!  
Processing: 3_Ronin.txt|Script file found|Chunks:6|.....|490 KP!  
Processing: 3_Snatch.txt|Script file found|Chunks:9|.....|700 KP!  
Processing: 4_Dumb and Dumber.txt|Script file found|Chunks:11|.....|879 KP!  
Processing: 4_Monty Pythons Holy Grail.txt|Script file found|Chunks:7|.....|574 KP!  
Processing: 4_The Hangover.txt|Script file found|Chunks:14|.....|947 KP!  
Processing: 4_Zoolander.txt|Script file found|Chunks:9|.....|843 KP!  
Processing: 5_2001 Space Odyssey.txt|Script file found|Chunks:5|.....|511 KP!  
Processing: 5_Alien.txt|Script file found|Chunks:5|.....|398 KP!  
Processing: 5_Blade Runner.txt|Script file found|Chunks:4|....|372 KP!  
Processing: 5_Star Wars.txt|Script file found|Chunks:10|.....|748 KP!
```



Label	Key Phrases
001	boyfriend girlfriend right time gonna flip pair of gloves nice time great time search of black gloves b
001	afraid I Walter Bees tomato Cold salmon minutes Hug wedding strawberries Pride things Annie wonderful t
001	kind of love love stories love fades unrequited love life happy Christmas Shakespeare God absolute wors
001	life great sex big Sheldon pie greatest sex good sex Sally gonna Wanna right mind New York happy person
002	Blair Witch old story woods slate old hermit Blair High School different story scary story creepiest st
002	Adam game room thing Help tape dead body Lawrence Gordon way home Diana kidneys Alison chains heart ma
002	animal Damn thing Jesus Christ Ooooooooo Scotty look stupidest thing Book old place Cheryl hell years o
002	Jack Torrance people winter caretaker winter sports Denver real good time good references little time h
003	Baby job gonna good kid Shh Sound good river of sin waiting getaway car time doc calls police hands whe
003	Good man years Angelo Pappas Good luck Good job Good moves hands good crimescene work good lab work dro
003	Vincent weapons man gonna job Country mission ambush case long way wheelchair transit Interrogation wor
003	Boris minutes gun ucking floor virgin stone nice story protection plane leave Michael Russian plane cr
004	fe wine Ha ha ha sir fe line Ragamuffin style fe say Lloyd Christmas white courtesy phone white zone wi
004	Britons Arthur castle of Camelot coconuts court of Camelot European swallow stone dead order master lor
004	DOUG Vegas PHIL love PHONE LINE DIALING Alan drive gonna car PHONE LINE BEEPS RINGING CELL PHONE RINGS
004	new look Blue Steel look new millennium Derek Zoolander Fashion Week fashion icon new afro lot of time
005	nice birthday party Looking weII nice present I'II days Heywood FIoyd WouId CIavius good fIight teII Mu
005	seconds time commercial ship systematised transmission Antarctica traffic control transmission of unknow

Step #2 Train a multi-classifier ML.NET model

```
C:\WS\Rom-Com\Rom-Com internal>dotnet script ModelBuilder.csx train "lets drive and spend christmas in the city and party"
Hello! We are going to do some ML now.
I am expecting a data file of a header row with a category in column 1 and a list of words following. A TSV
I will then learn which words go against which category and you can test me
Training the model...
Saving model...
True
Loading model...

We have 5 scores to look at
-----
RomCominess |=====| 0.828126311
Horroness   |=====| 0.090081491
Heistiness  |===|    0.031740647
Comediness  |==|     0.024050022
Scifiness   |==|     0.026001520
-----
I think 'lets drive and spend christmas in the city and party' sounds most like a 'Rom Com'
-----
```

Step #3 Deploy behind dotnet core WebAPI

POST ▼ https://localhost:5001/predict Send Save ▼

Params Authorization Headers (9) **Body** Pre-request Script Tests Settings Cookies Code

● none ● form-data ● x-www-form-urlencoded ● raw ● binary ● GraphQL BETA JSON ▼ Beautify

```
1 ["And then, suddenly, it's the weekend, and from break of day hundreds of stalls appear out of nowhere, filling Portobello Road, right up to Notting Hill Gate. And wherever you look thousands of people are buying millions of antiques, some genuine and some... not quite so genuine. And what's great is that lots of friends have ended up in this part of London. That's Tony, for example, architect turned chef, who recently invested all the money he ever earned in a new restaurant. And so, this is where I spend my days and years... in this small village in the middle of the city in a house with a blue door... that my wife and I bought together before she left me for a man... who looked exactly like Harrison Ford. And where I lead a strange half-life with a lodger called-- Spike! You couldn't help me with an incredible important decision, could you? Is this important in comparison to, let's say, whether they should cancel Third World debt? That's right. I am at last going out on a date with the great Janine, and I just wanna be sure I've picked the right T-shirt. - What are the choices? - Well, wait for it. First there's this one. Cool, huh? Yeah, it might make it hard to strike a really romantic note. Point taken. Don't despair. If it's romance we're looking for, I believe I have just the thing. Yeah, well, there again, she might not think you had true love on your mind. Right. Just one more. True love here I come. Well, yeah. Yeah, that's-- that's, um, perfect. Great. Thanks. - Wish me luck. - Good luck. And so it was just another hopeless Wednesday, as I walked the thousand yards through the market to work, never suspecting that this was the day that was gonna change my life forever. This is work, by the way, my little travel bookshop."]
```

Body Cookies Headers (4) Test Results Status: 200 OK Time: 1651ms Size: 363 B Save Response ▼

Pretty Raw Preview Visualize BETA Text ▼ ≡

```
1 [{"genre":"RomCom","probability":0.527722}, {"genre":"Horror","probability":0.38999084}, {"genre":"Heist","probability":0.010471797}, {"genre":"Comedy","probability":0.06425919}, {"genre":"Sci-Fi","probability":0.0075562103}]
```

Home > romcomnotromcom - Deployment Center

romcomnotromcom - Deployment Center

App Service Directory: Microsoft

Search (Ctrl+/)

Overview
Activity log
Access control (IAM)
Tags
Diagnose and solve problems
Security

Deployment
Quickstart
Deployment slots
Deployment Center





Settings
Configuration
Authentication / Authorization
Application Insights
Identity
Backups
Custom domains
TLS/SSL settings
Networking
Scale in (App Service plan)

Deployment Center

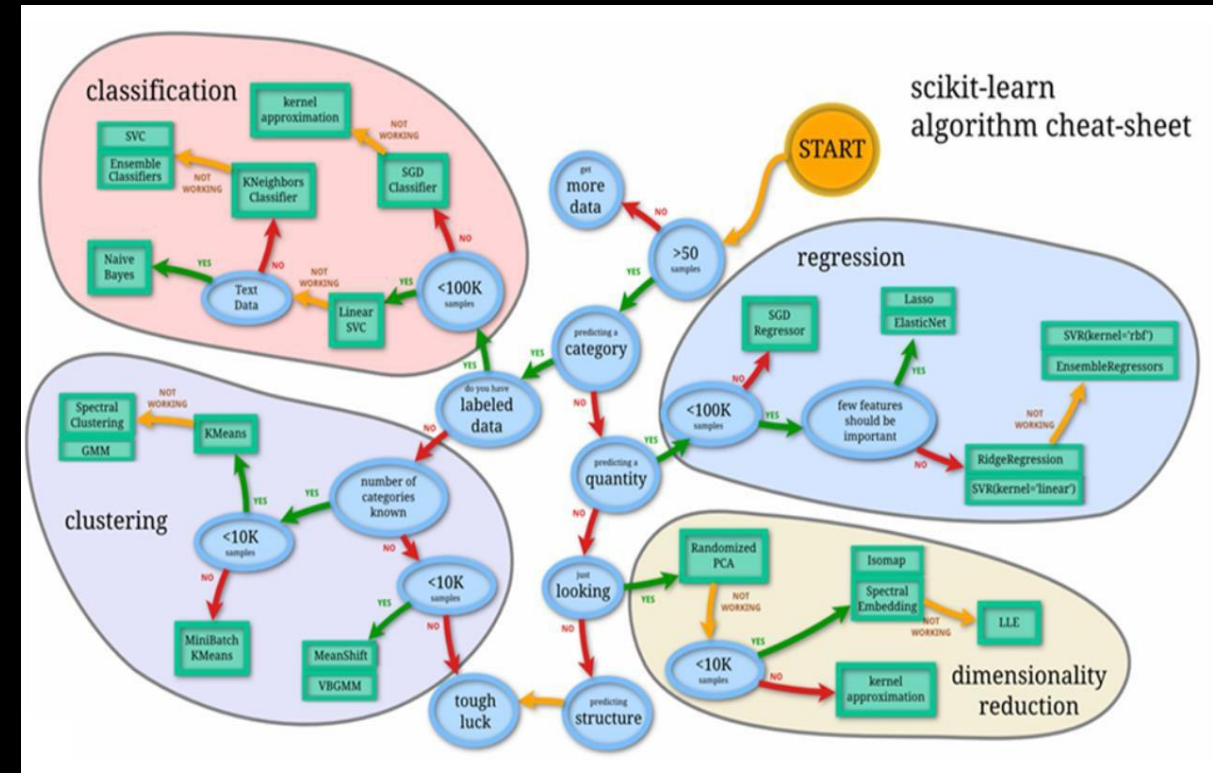
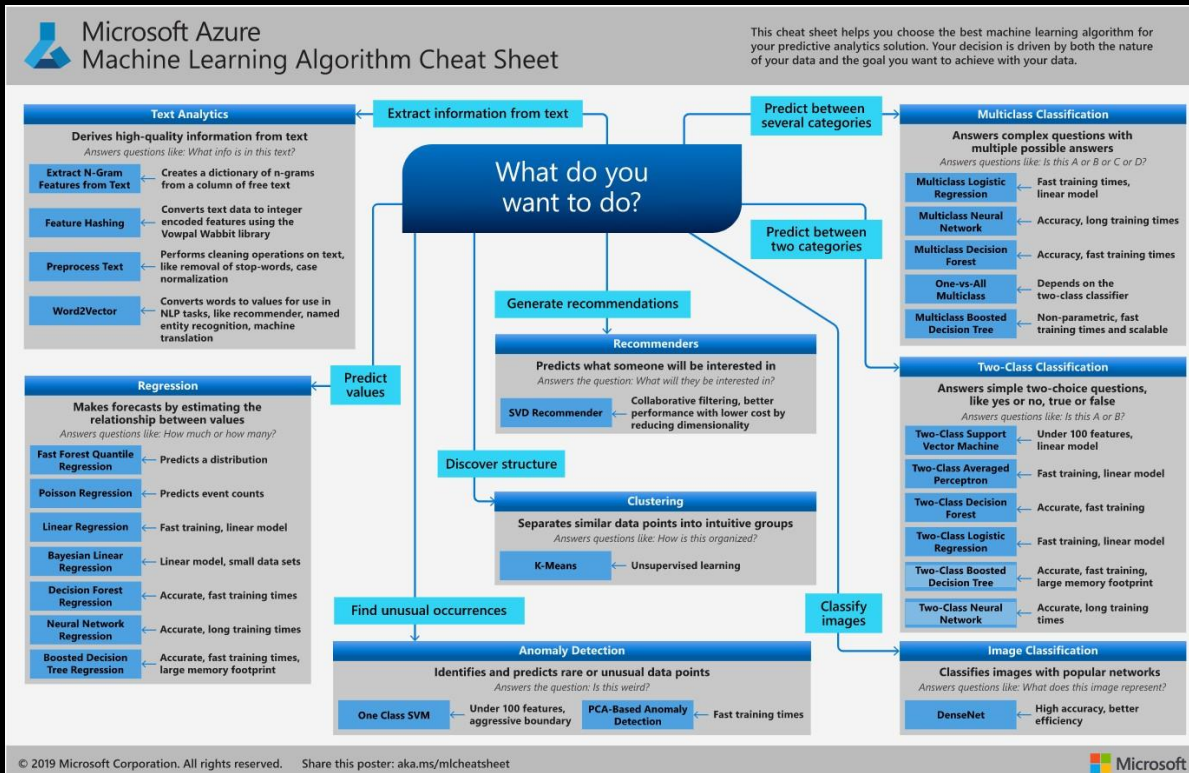
App Service Deployment Center enables you to choose the location of your code as well as options for build and deployment to the cloud. [Learn more](#)

1 SOURCE CONTROL 2 BUILD PROVIDER 3 CONFIGURE

Continuous Deployment (CI / CD)

 Azure Repos Configure continuous integration with an Azure Repo, part of Azure DevOps Services (formerly known as VSTS).	 GitHub Configure continuous integration with a GitHub repo. davidgristwood	 Bitbucket Configure continuous integration with a Bitbucket repo. Not Authorized
 Local Git Deploy from a local Git repo.		

The need for AutoML



ML.NET AutoML

AutoML performs automatic model selection and training

You supply machine learning task and supply a dataset, it outputs:

- a model file that can be loaded into your prediction application
- application code to make predictions
- the source code used for feature selection and model training (to understand the model)

Available in 3 form factors:

- ML.NET CLI
- ML.NET Model Builder
- ML.NET AUTO API

MLNET CLI

mlnet auto-train

--task **binary-classification**
--dataset "**yelp_labelled.txt**"
--label-column-index 1
--has-header false
--max-exploration-time 10

Current supported ML Tasks

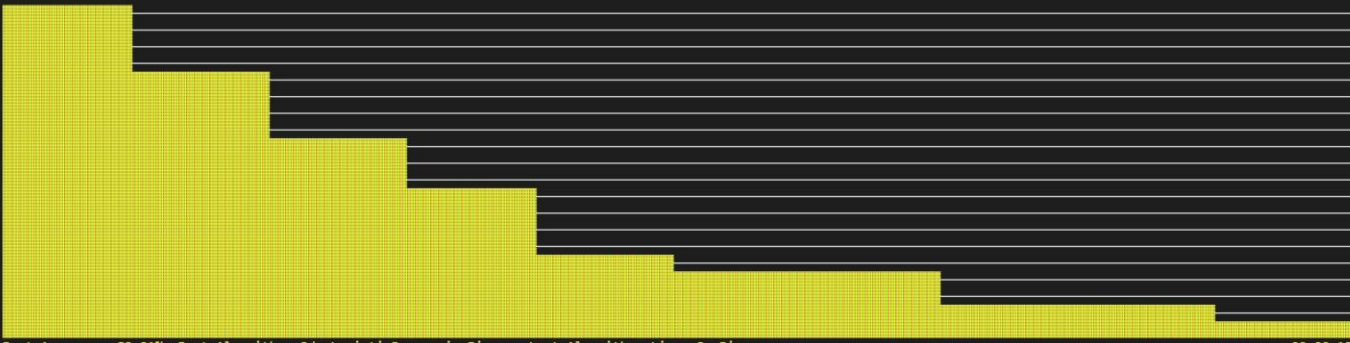
- **binary-classification**
- **multiclass-classification**
- **Regression**

Future:

- recommendation
- anomaly-detection
- clustering

```
yelp_labelled.txt
1 Wow... Loved this place. 1
2 Crust is not good. 0
3 Not tasty and the texture was just nasty. 0
4 Stopped by during the late May bank holiday off Rick Steve recommendation and loved it. 1
5 The selection on the menu was great and so were the prices. 1
6 Now I am getting angry and I want my damn pho. 0
7 Honestly it didn't taste THAT fresh.) 0
8 The potatoes were like rubber and you could tell they had been made up ahead of time being kept under a warmer. 0
```

```
PS C:\WS\MLNET\CLI-Demo\Nov2019> mlnet auto-train --task binary-classification --dataset "yelp_labelled.txt" --label-column-index 1 --has-header false --max-exploration-time 10
Exploring multiple ML algorithms and settings to find you the best model for ML task: binary-classification
For further learning check: https://aka.ms/mlnet-cli
```



```
Best Accuracy: 83.91%, Best Algorithm: SdcaLogisticRegressionBinary, Last Algorithm: LinearSvmBinary 00:00:13
```

```
=====Experiment Results=====
|
|-----Summary-----
|
| ML Task: binary-classification
| Dataset: yelp_labelled.txt
| Label : Label
| Total experiment time : 15.05 Secs
| Total number of models explored: 5
|
|-----Top 5 models explored-----
|
| Trainer                                     Accuracy   AUC   AUPRC   F1-score   Duration #Iteration
|-----|-----|-----|-----|-----|-----|
| 1 SdcaLogisticRegressionBinary              0.8391    0.9313  0.9284   0.8511      0.8         2
| 2 AveragedPerceptronBinary                  0.8276    0.8864  0.8645   0.8387      1.6         1
| 3 LightGbmBinary                            0.8276    0.8874  0.8776   0.8454      1.8         3
| 4 LinearSvmBinary                           0.8046    0.9022  0.8877   0.8046      0.9         5
| 5 SymbolicSgdLogisticRegressionBinary       0.7811    0.8192  0.8137   0.7451      1.0         4
|
Generated trained model for consumption: C:\WS\MLNET\CLI-Demo\Nov2019\SampleBinaryClassification\SampleBinaryClassification.Model\MLModel.zip
Generated C# code for model consumption: C:\WS\MLNET\CLI-Demo\Nov2019\SampleBinaryClassification\SampleBinaryClassification.ConsoleApp
Check out log file for more information: C:\WS\MLNET\CLI-Demo\Nov2019\SampleBinaryClassification\logs\debug_log.txt
```

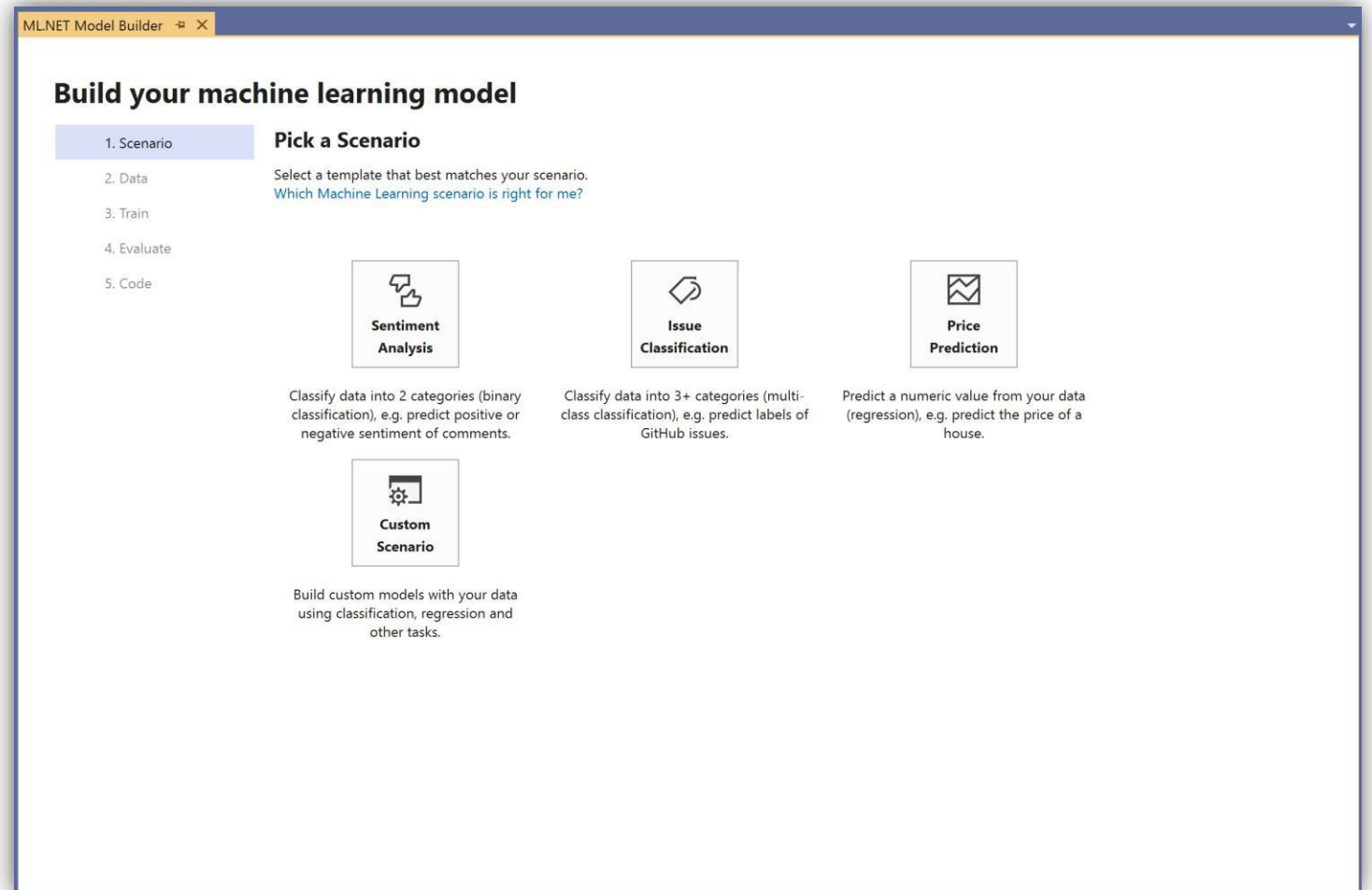
ML.NET Model Builder

Approachable machine learning in Visual Studio

- A simple UI to easily build custom ML models with Automated ML
- Load from files and databases
- Generate code for training and consumption
- Run everything local

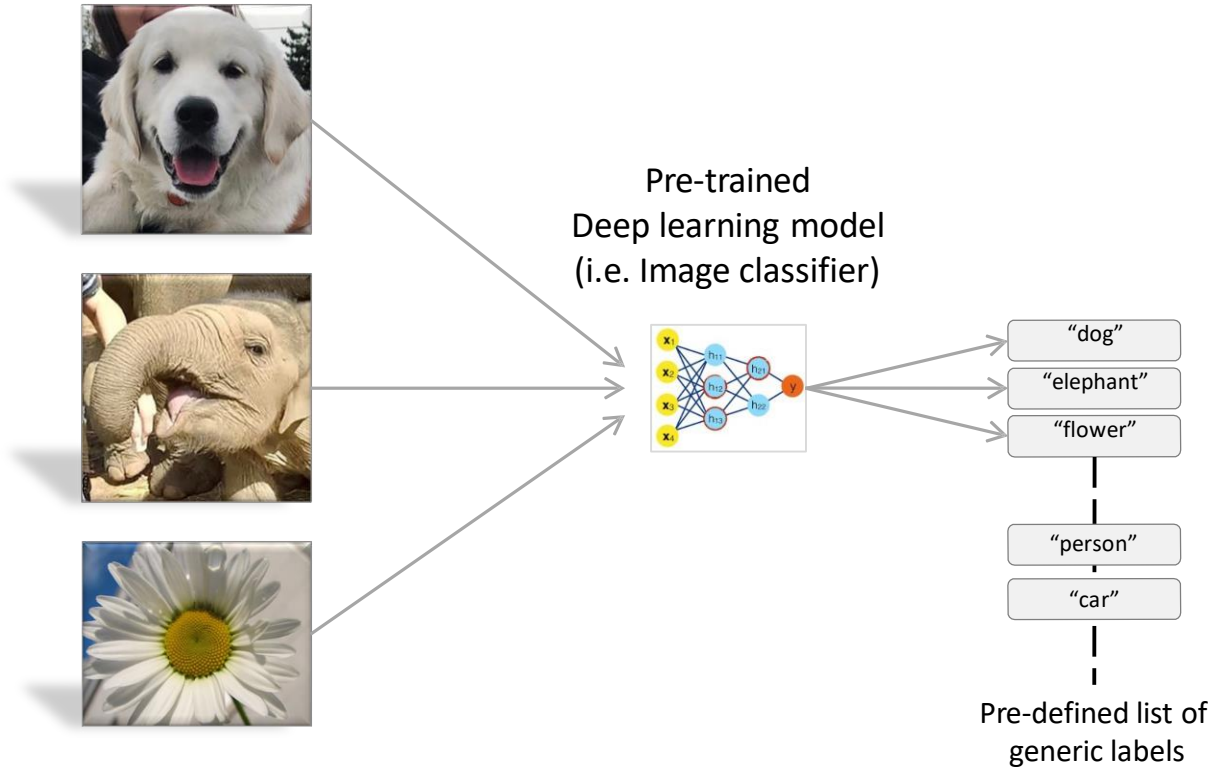
Download VS vsix:

<http://aka.ms/mlnetmodelbuilder>



Consuming pre-trained deep learning models with ML.NET

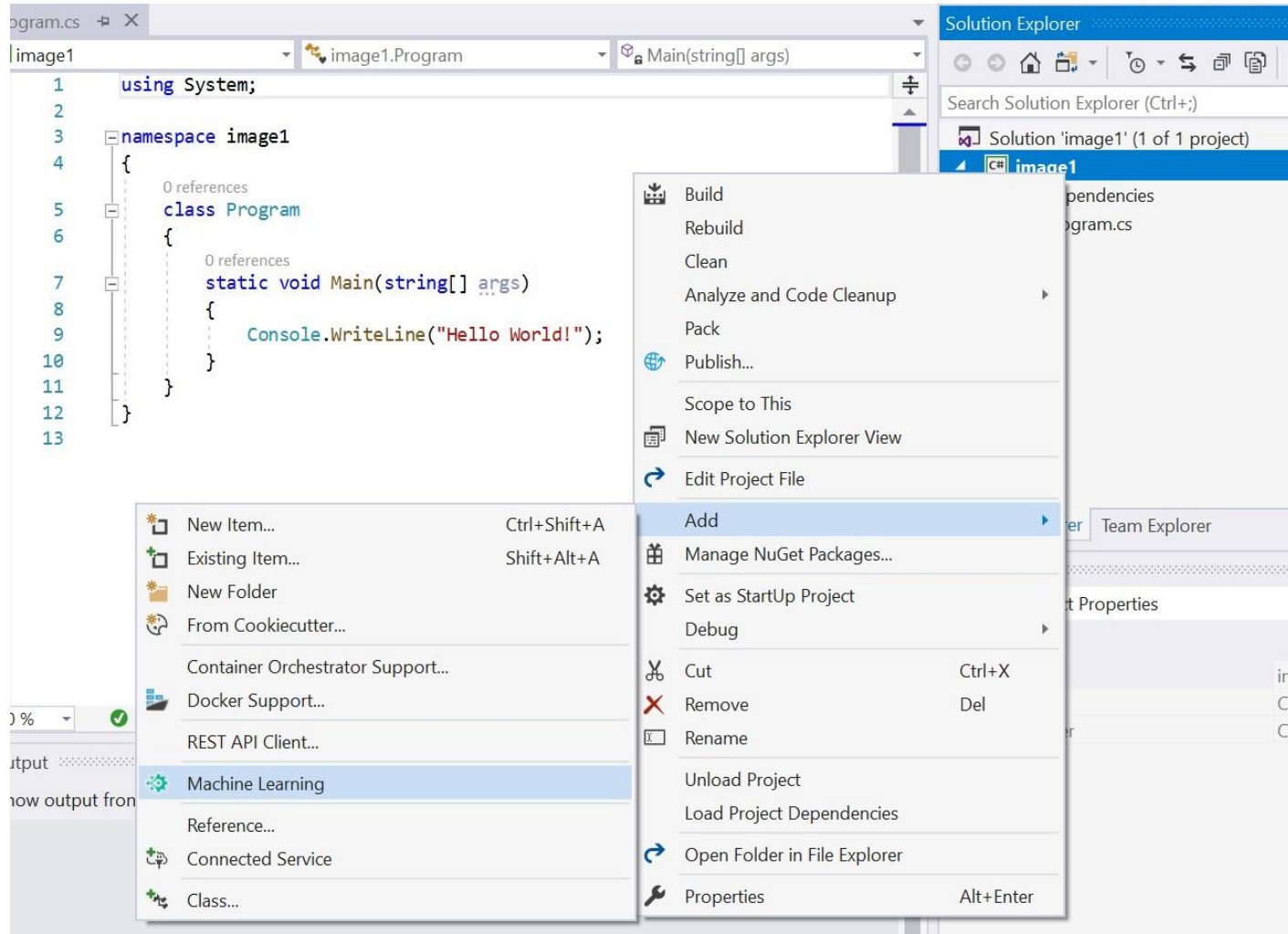
Scenario: **Image classifier** (Consuming the model)




Examples of pre-trained models (Image classifiers):

- Google **Inception v3**, **NASNet**
- Microsoft **ResNet**
- Oxford **VGG** Model, etc.

Add model to project



#1 Chose scenario

ML.NET Model Builder  taxi-fare-train.csv Program.cs

1. Scenario

2. Data


3. Train

4. Evaluate

5. Code


Pick a Scenario

Select a template that best matches your scenario.
[Which Machine Learning scenario is right for me?](#)




Sentiment Analysis

Classify data into 2 categories (binary classification), e.g. predict positive or negative sentiment of comments.



Issue Classification

Classify data into 3+ categories (multi-class classification), e.g. predict labels of GitHub issues.



Price Prediction

Predict a numeric value from your data (regression), e.g. predict the price of a house.





Image Classification

Classify images into 2+ categories (image-classification) e.g. predict type of flowers.



Custom Scenario

Build custom models with your data using classification, regression and other tasks.

#2 Add data

✓ 1. Scenario

✓ 2. Data

3. Train

4. Evaluate

5. Code

Add data

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

Input

Select the folder which contains all your images. This folder should organize your photos into separate labeled sub-folders.

Select a folder:

C:\WS\MLNET\ModelBuilder\ImageClassification\ca ...

Supported file formats: .png, .jpg, .jpeg, .gif.

Example folder structure:

```
Images
├── Label 1
│   ├── Image 1
│   └── Image 2
├── Label 2
└── Label 3
```

Data Preview

Total images 30. Showing 12/15.

Animals:

Cat (15)

Dog (15)



#3 Train model

✓ 1. Scenario

✓ 2. Data

✓ 3. Train

4. Evaluate

5. Code

Train

Model Builder automatically sets the training time for image classification based on the size of your dataset.

Start training

Progress

Status: ✓ Training complete

Accuracy: 50%

Algorithm: DNN+ResNet50

Next Step: [Evaluate](#)

Output

Show output from: Machine Learning

[Source=ImageClassificationTrainer; EmptyDataView; Cursor, Kind=Trace] Channel started
[Source=ImageClassificationTrainer; EmptyDataView; Cursor, Kind=Trace] Channel finished. Elapsed 00:00:00.0003814.
[Source=ImageClassificationTrainer; EmptyDataView; Cursor, Kind=Trace] Channel disposed
[Source=ImageClassificationTrainer; BinarySaver; Write, Kind=Trace] Channel started
[Source=ImageClassificationTrainer; EmptyDataView; Cursor, Kind=Trace] Channel started
[Source=ImageClassificationTrainer; EmptyDataView; Cursor, Kind=Trace] Channel finished. Elapsed 00:00:00.0002885.
[Source=ImageClassificationTrainer; EmptyDataView; Cursor, Kind=Trace] Channel disposed
[Source=ImageClassificationTrainer; BinarySaver; Write, Kind=Trace] Channel finished. Elapsed 00:00:00.0037231.
[Source=ImageClassificationTrainer; BinarySaver; Write, Kind=Trace] Channel disposed
[Source=ImageClassificationTrainer; BinarySaver; Saving, Kind=Trace] Channel finished. Elapsed 00:00:00.0081508.
[Source=ImageClassificationTrainer; BinarySaver; Saving, Kind=Trace] Channel disposed
[Source=ImageClassificationTrainer; ImageClassificationTrainer, Kind=Trace] Channel started
|1 ImageClassification 0.5000 0.5000 607.2 1 |

=====Experiment Results=====

| Summary |

| ML Task: multiclass-classification |

| Dataset: C:\Users\davidgri.000\AppData\Local\Temp\4748fcf9-1a67-426f-9adf-682fd634a583.tsv |

| Label : Label |

| Total experiment time : 607.2 Secs |

| Total number of models explored: 1 |

| Top 1 models explored |

| Trainer MicroAccuracy MacroAccuracy Duration #Iteration |

| 1 ImageClassification 0.5000 0.5000 607.2 1 |

Code Generated

#4 Evaluate model

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

Output

Overview Details

ML Task: image-classification
Training Time: 578.6 seconds
Models Explored (Total): 1 | [View Top 1 model explored](#)

Overall accuracy:
100%

Try your model



Results:

Dog	99%
Cat	1%

Predict

[Test another image](#)

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

Output

Overview Details

ML Task: image-classification
Training Time: 578.6 seconds
Models Explored (Total): 1 | [View Top 1 model explored](#)

Overall accuracy:
100%

Try your model



Results:

Dog	87%
Cat	13%

Predict

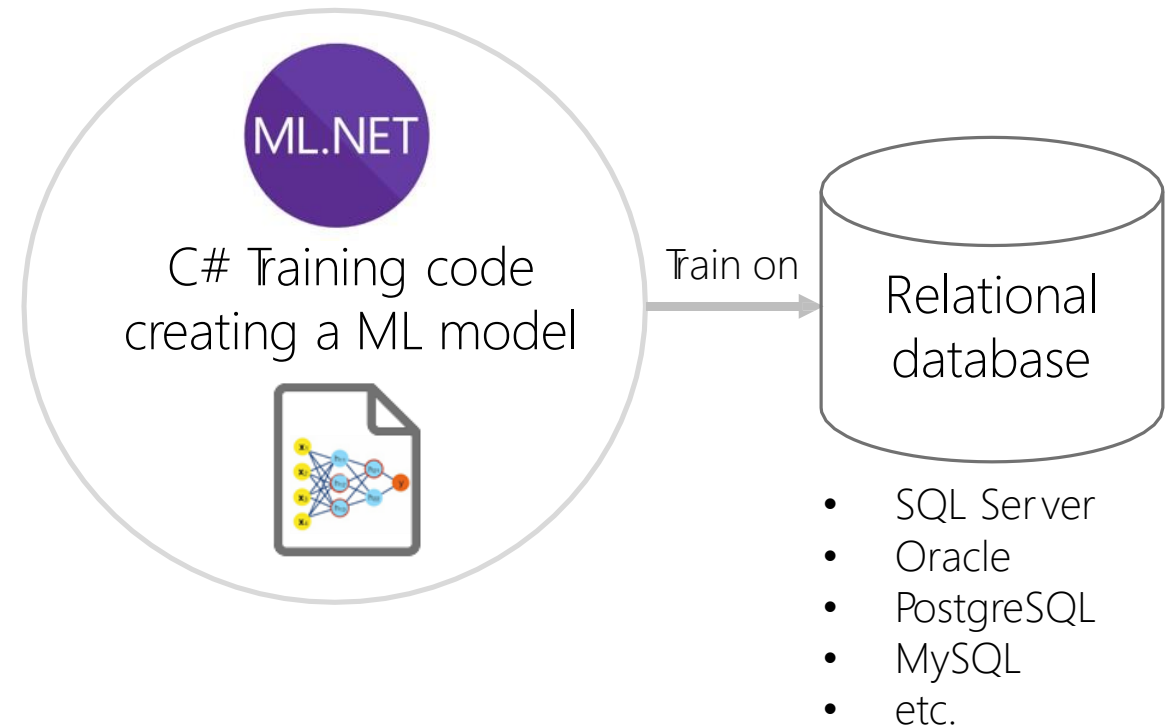
[Test another image](#)

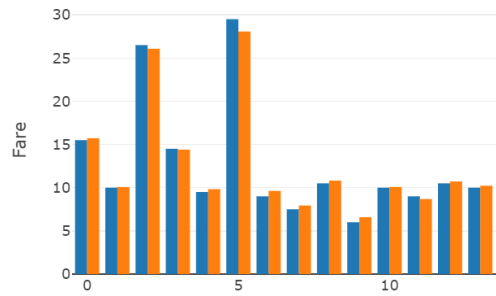
Latest updates...

Database Loader

Enabled scenarios:

- Training directly against relational databases.
- Simple and out-of-the-box code
- Supports any RDBMS supported by System.Data
- v1.4-preview release





The Jupyter Notebook

The Jupyter Notebook is an open-source web application that allows you to create and share documents that contain live code, equations, visualizations and narrative text. Uses include: data cleaning and transformation, numerical simulation, statistical modeling, data visualization, machine learning, and much more.

```

jupyter MLNET-Jupyter-Demo Last Checkpoint: 23 minutes ago (autosaved)

File Edit View Insert Cell Kernel Widgets Help

Load datasets into IDataView and display

In [4]: display(h1("Code for loading the data into IDataViews: training dat
MLContext mlContext = new MLContext(seed: 0);

string TrainDataPath = "./taxi-fare-train.csv";
string TestDataPath = "./taxi-fare-test.csv";

IDataView trainDataView = mlContext.Data.LoadFromTextFile<TaxiTrip>
IDataView testDataView = mlContext.Data.LoadFromTextFile<TaxiTrip>('

```



ML.NET and C# in Jupyter Notebooks!

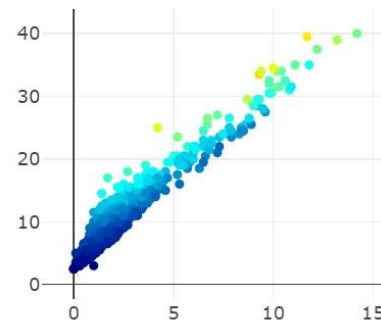
jupyter MLNET-Jupyter-Demo Last Checkpoint: 24 minutes ago (autosaved)

File Edit View Insert Cell Kernel Widgets Help

Showing a few rows from training DataView:

DataView: Showing 5 rows with the columns

Index	VendorId	RateCode	PassengerCount	TripTime	TripDistance	PaymentType
0	CMT	1	1	1271	3.8	CRD
1	CMT	1	1	474	1.5	CRD
2	CMT	1	1	637	1.4	CRD
3	CMT	1	1	181	0.6	CSH
4	CMT	1	1	661	1.1	CRD



Roadmap ahead

- Scale-out on [Azure](#) for training and consumption
- Improve [tooling](#) in Visual Studio and .NET ([Model Builder](#) & [CLI](#))
 - Improve [AutoML.NET](#) experience for all ML scenarios
- Object Detection training (Deep Learning based)
- Text Analysis support (Deep Learning based, i.e. BERT)
- ARM / full ONNX support (Enablers for Xamarin & IoT scenarios)

Bibliography



Get started at <http://dot.net/ml>



Try the samples at <http://aka.ms/mlnetsamples>



Read the docs at <http://aka.ms/mlnetdocs>



Watch ML.NET videos at <https://aka.ms/mlnetyoutube>



Request features or contribute at <http://aka.ms/mlnet>