

Making an Art Critic using Machine Learning

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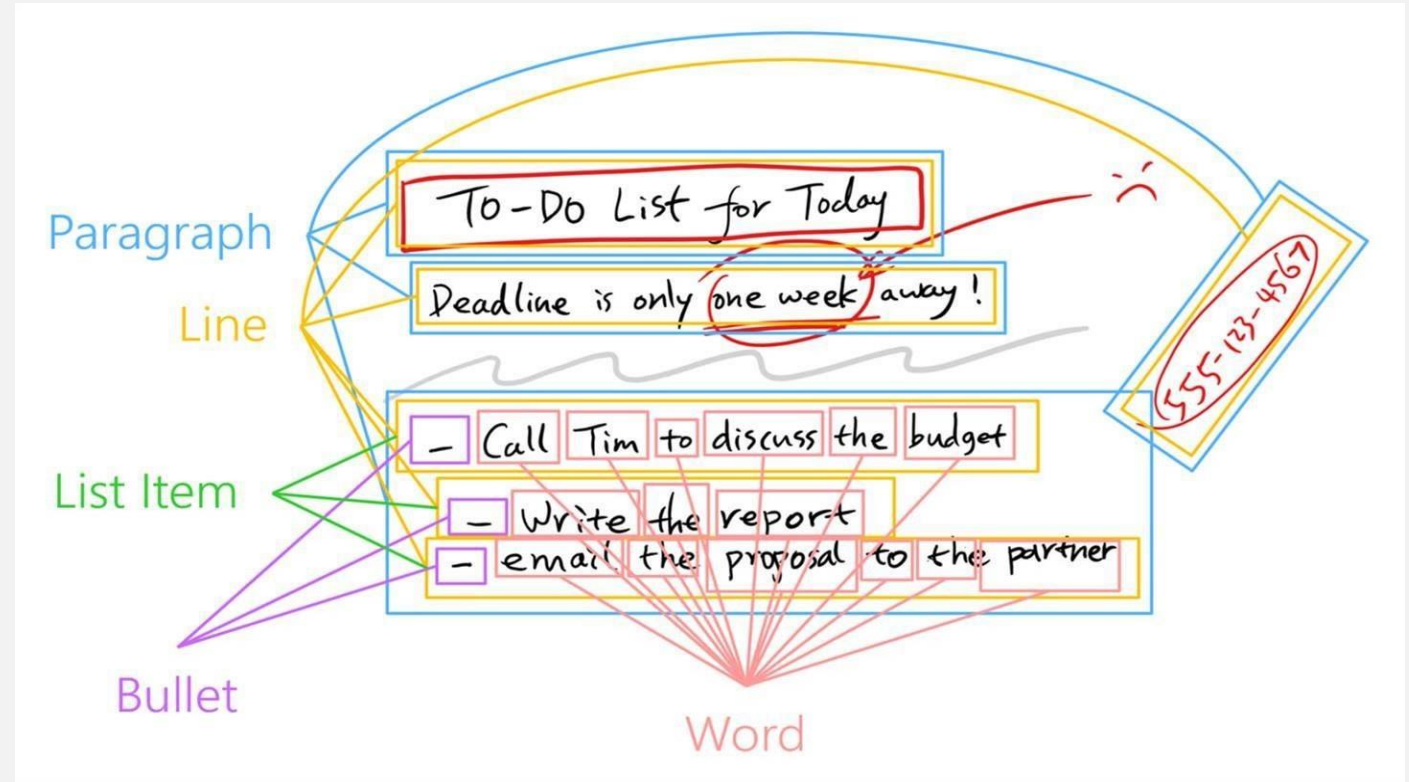
Machine Learning Makes Hard Things Easy

Computers tend to be good at things humans are bad at, and vice versa

Good at chess

Bad at recognising people

Machine Learning (ML) can redress the balance and make the computer good at stuff in a useful way



Machine Learning Makes Hard Things Easy

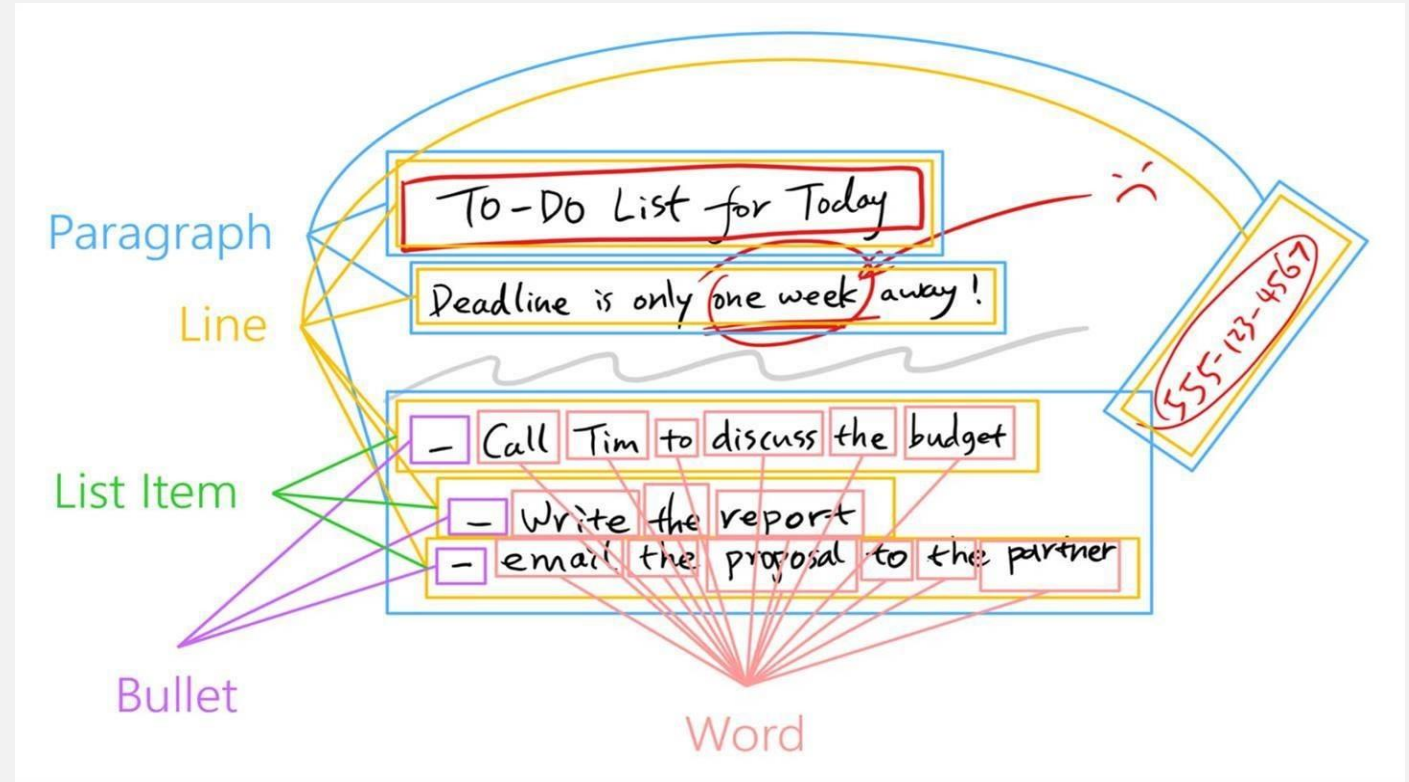
Computers tend to be good at things humans are bad at, and vice versa

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Machine Learning (ML) can redress the balance and make the computer good at stuff in a useful way

Even if we don't know precisely how it works.....



Building an Art Critic

I'm not very good with art

But I know a good picture when I see it

With that in mind, the starting point for a computerised art critic should be something that can recognise different objects

So, lets build one using Machine Learning



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What we are going to do...

Use training data to build model for image recognition

Incorporate the training data in a Universal Windows Application

Deploy the application and run it on the target device

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Drop the model into Visual Studio and see how our program can bind to it

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 - Drop the model into Visual Studio and see how our program can bind to it

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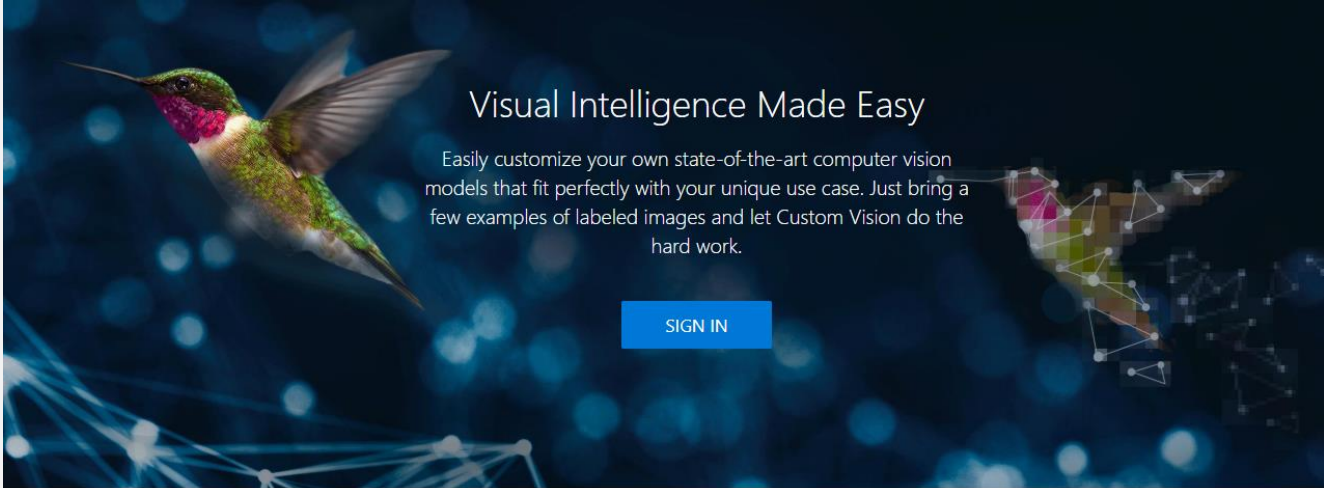
 - Build and run the application

Microsoft Custom Vision

You can create your own models in the cloud using Microsoft Custom Vision

- Sign in using your Azure account
- Upload images
- Build the model
- Download it for use in your application

www.customvision.ai

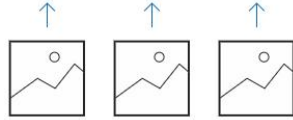


The landing page features a hummingbird on the left and a wireframe hummingbird on the right, set against a dark blue background with bokeh light effects. The text 'Visual Intelligence Made Easy' is prominently displayed, followed by a description of the service and a 'SIGN IN' button.

Visual Intelligence Made Easy


Easily customize your own state-of-the-art computer vision models that fit perfectly with your unique use case. Just bring a few examples of labeled images and let Custom Vision do the hard work.

[SIGN IN](#)




Upload Images

Bring your own labeled images, or use Custom Vision to quickly add tags to any unlabeled images.



Train

Use your labeled images to teach Custom Vision the concepts you care about.



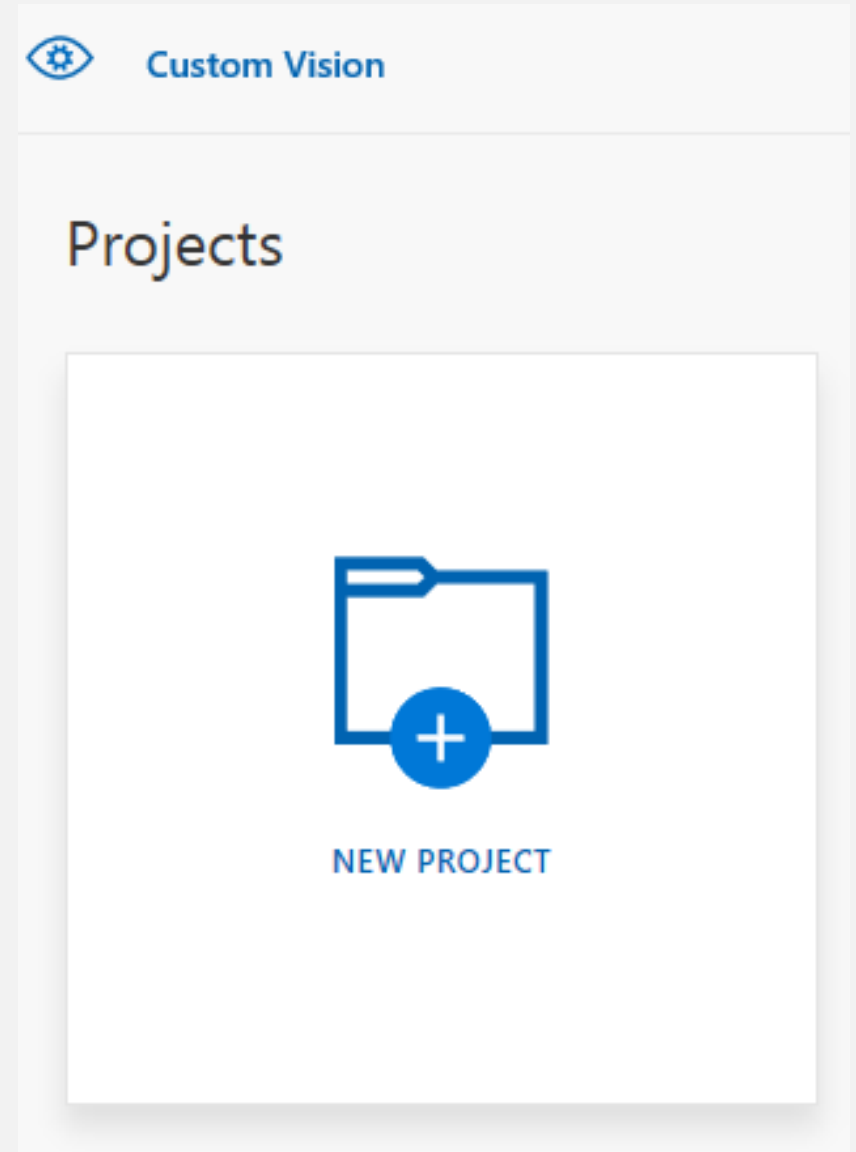
Evaluate

Use simple REST API calls to quickly tag images with your new custom computer vision model.

Custom Vision Project

In demo mode you can create two projects with a limited number of figures and classification types

Plenty to get started



Make a new project

These are the properties of a new project

Create new project

Name*

Art Critic

Description

This project can recognise simple objects

Project Types ⓘ

☒ Classification

☐ Object Detection

Classification Types ⓘ

☐ Multilabel (Multiple tags per image)

☒ Multiclass (Single tag per image)

Domains ⓘ

☐ General

☐ Food

☐ Landmarks

☐ Retail

☐ Adult

☒ General (compact)

☐ Food (compact)

☐ Landmarks (compact)

☐ Retail (compact)

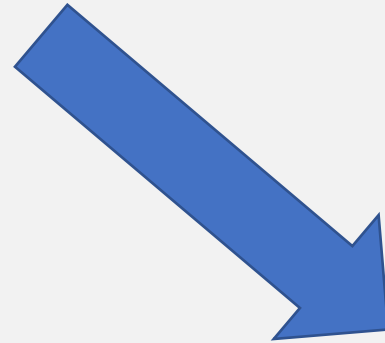
Cancel

Create project

Make a new project

These are the properties of a new project

Only compact models can be exported



Create new project

Name*

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Create project

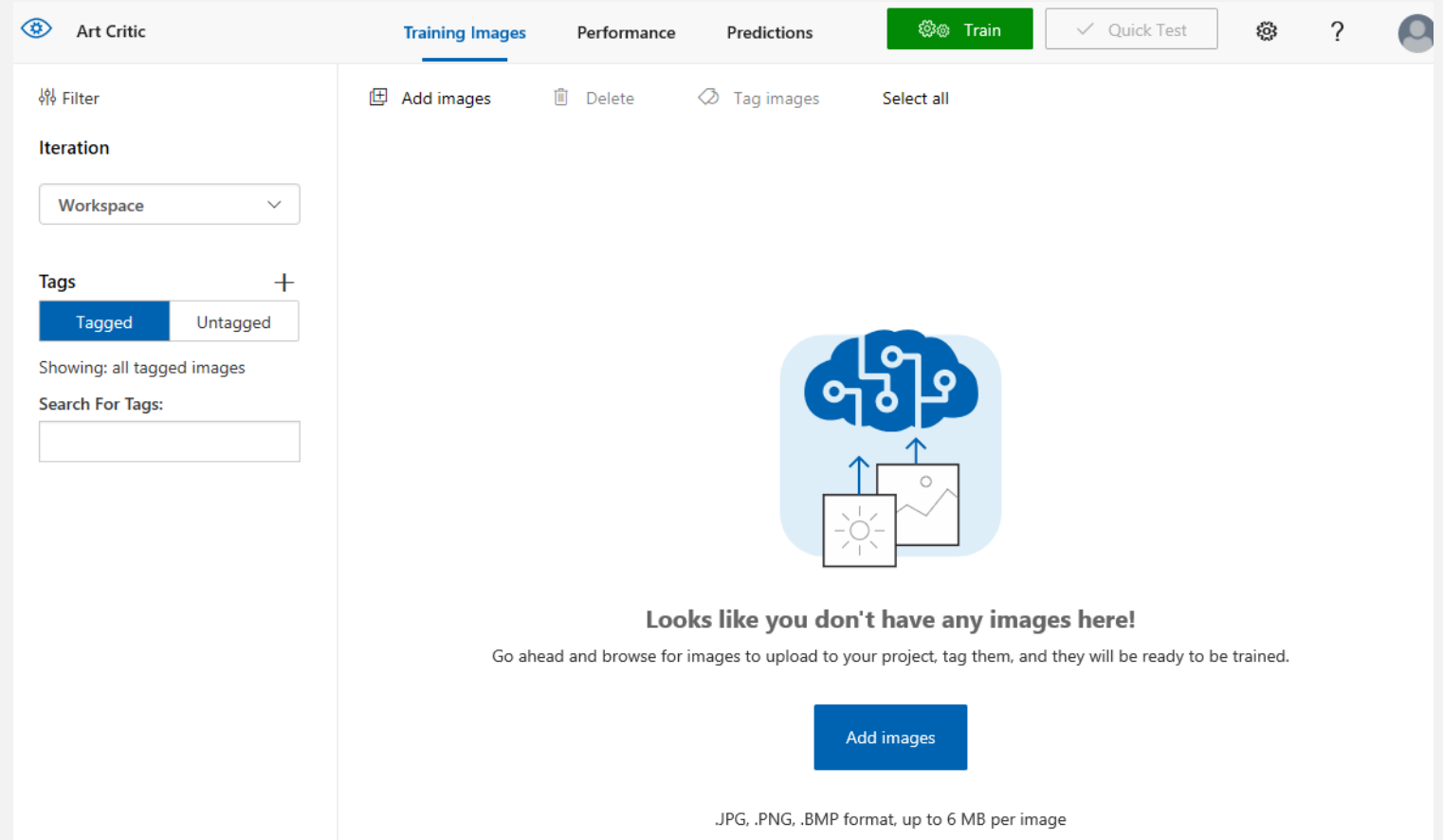
Images

Now that we have our project we can add some images

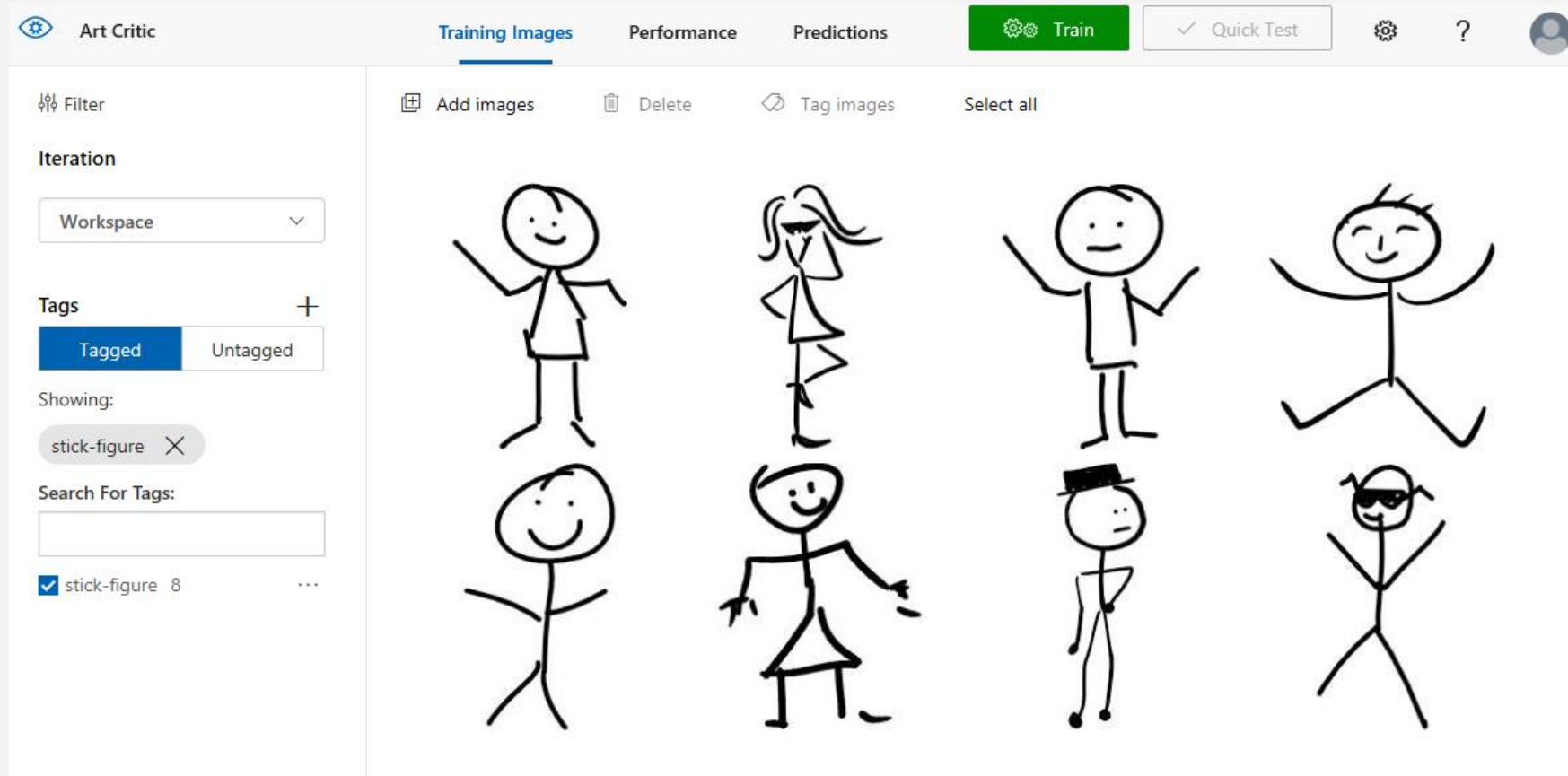
These can be in formats such as png or jpeg

Note that high resolution might not be your friend

The larger the image, the more data will have to be processed



Add some images

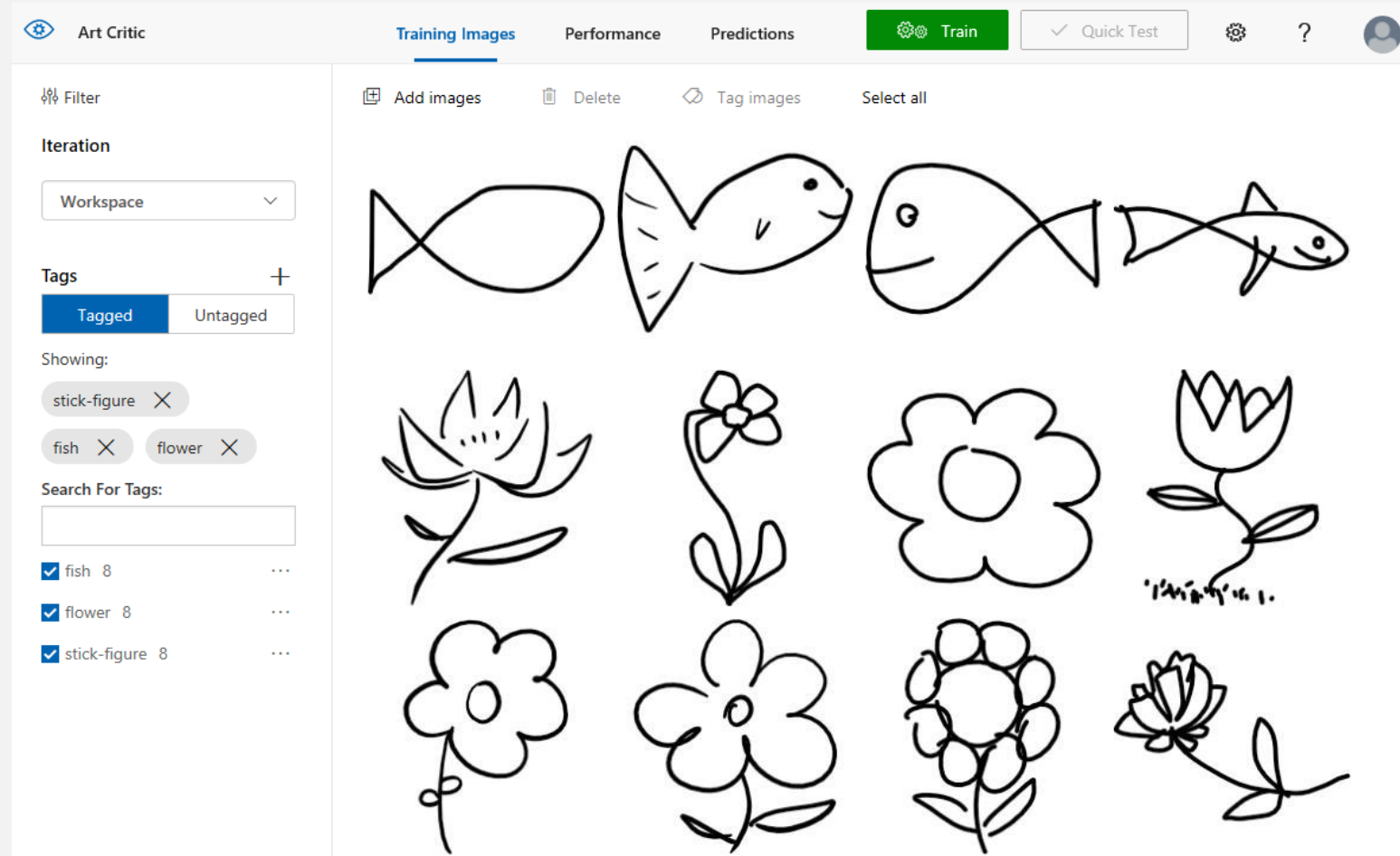


Choose your images carefully....

All images loaded

These are all the images loaded

Now we need to use this data to train a model

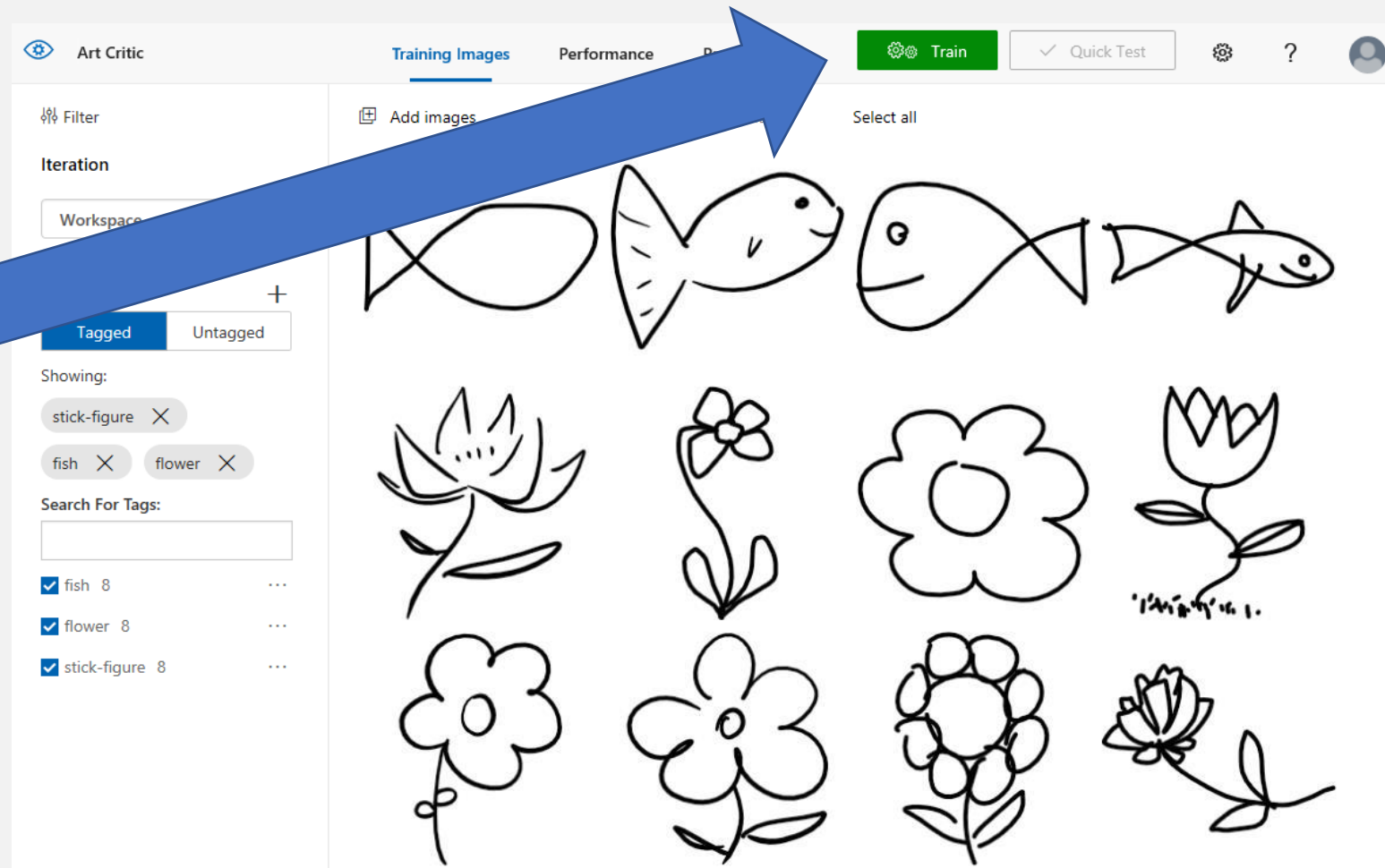


All images loaded

These are all the images loaded

Now we need to use this data to
train a model

Press the Train button to start



Training results

You get a report of how successful the training was

Precision:

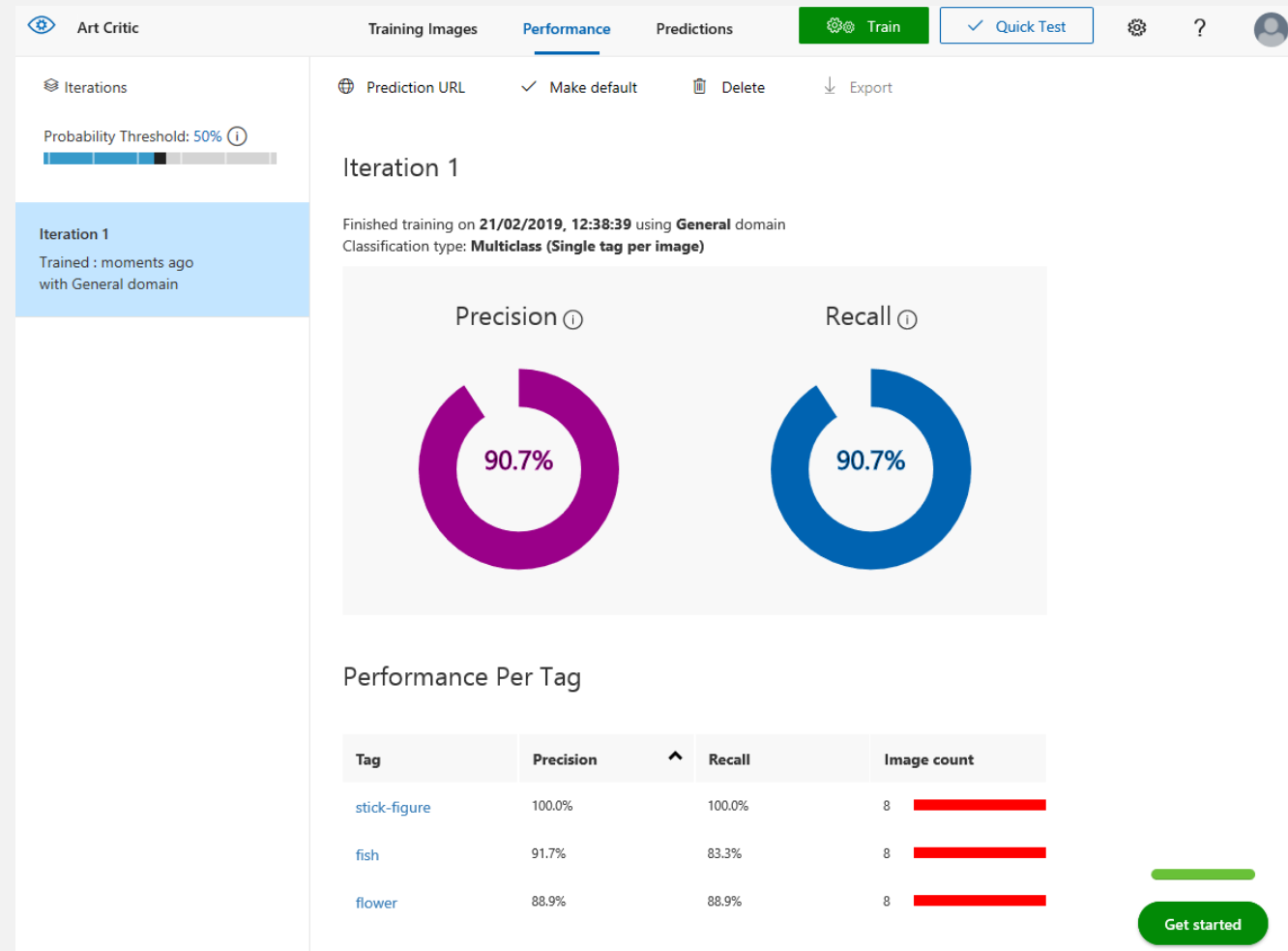
A measure of the likelihood of a tag being predicted correctly

Recall:

Out of the tags that should be predicted correctly, what percentage did the model find

You can use this to tweak your training data

You can also test your model against data with the Quick Test button



Exporting the model

We now have a model which can be used in applications to allow them to categorise an image based on the training data

We need to export the model to a file that we can incorporate in our application

Choose your platform



CoreML

iOS 11



TensorFlow

Android



ONNX

Windows ML



Dockerfile

Azure IoT Edge, Azure Functions,
AzureML

Exporting to ONNX

We are going to use the ONNX format

Version 1.2

Visual Studio can import files of this format

Choose your platform

ONNX

ONNX

ONNX1.2

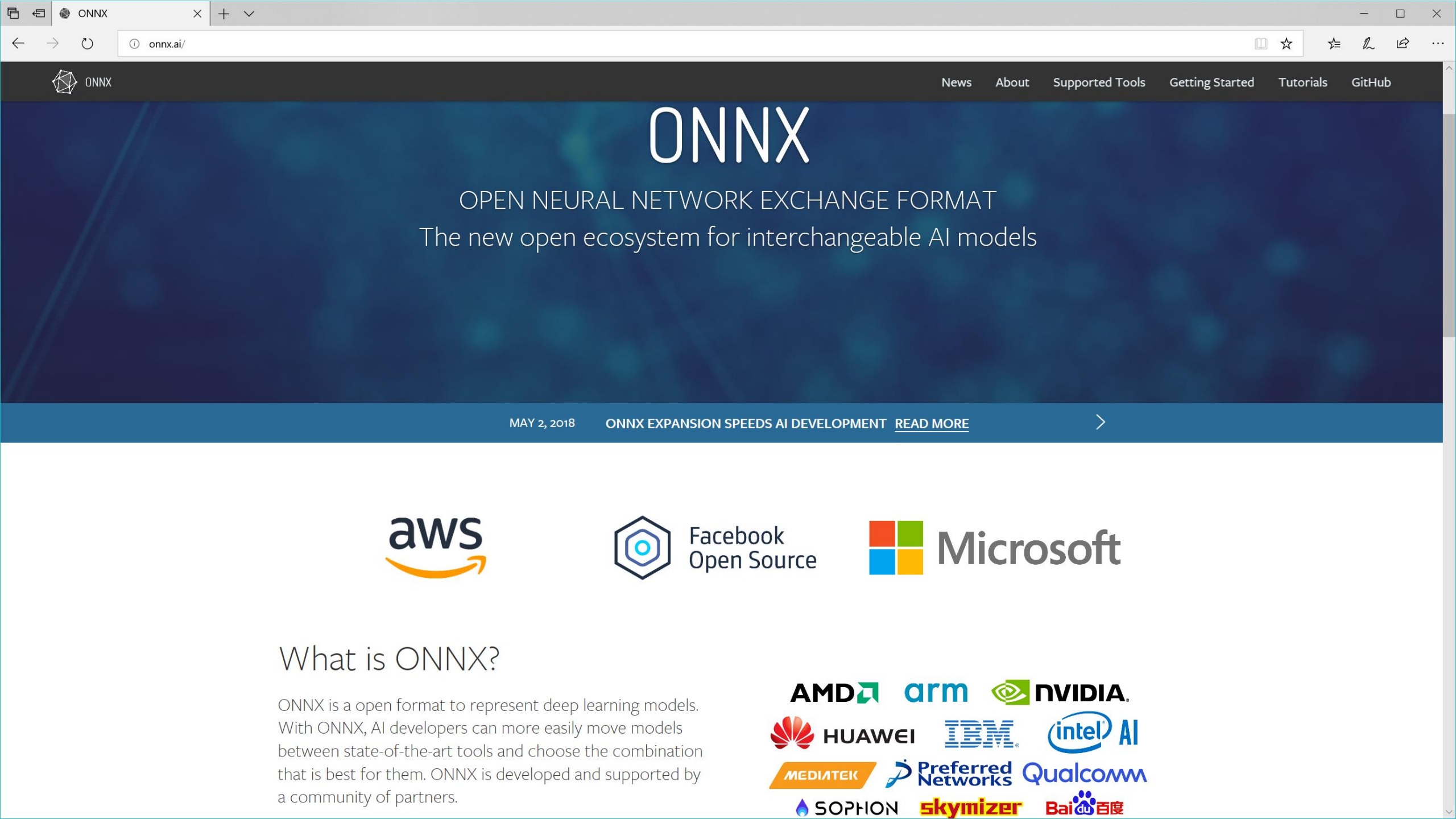
▼

ONNX1.2 only works for Windows 10 Build 17738 or higher

Export

How to use?

[ONNX documentation](#)
[Sample code ONNX 1.2](#)
[Licenses](#)



ONNX

News

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Supported Tools

Getting Started

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GitHub

ONNX

OPEN NEURAL NETWORK EXCHANGE FORMAT

The new open ecosystem for interchangeable AI models

MAY 2, 2018

ONNX EXPANSION SPEEDS AI DEVELOPMENT [READ MORE](#)



Facebook
Open Source



Microsoft

What is ONNX?

ONNX is a open format to represent deep learning models. With ONNX, AI developers can more easily move models between state-of-the-art tools and choose the combination that is best for them. ONNX is developed and supported by a community of partners.

AMD  arm  NVIDIA



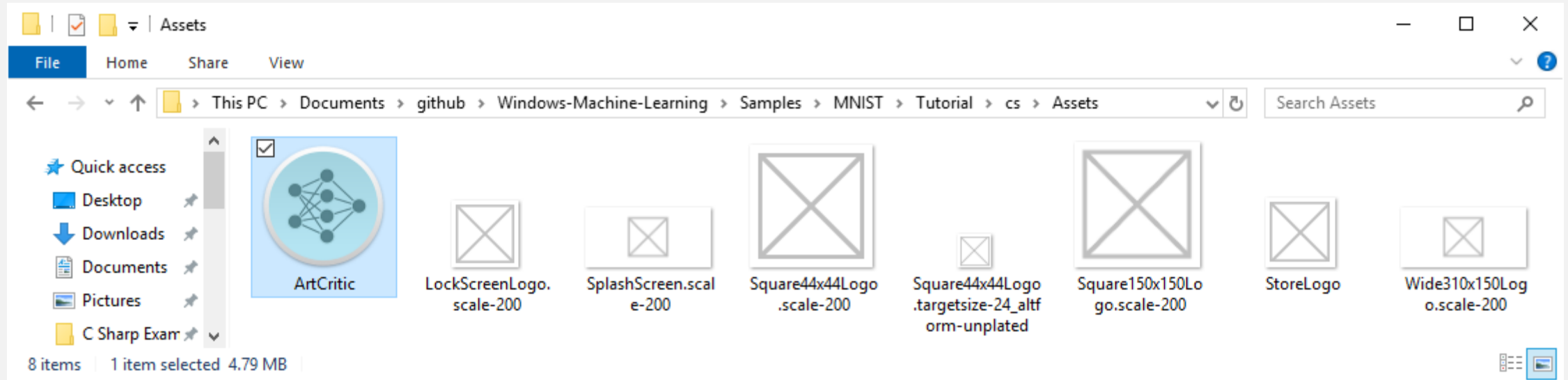
HUAWEI



SOPHION



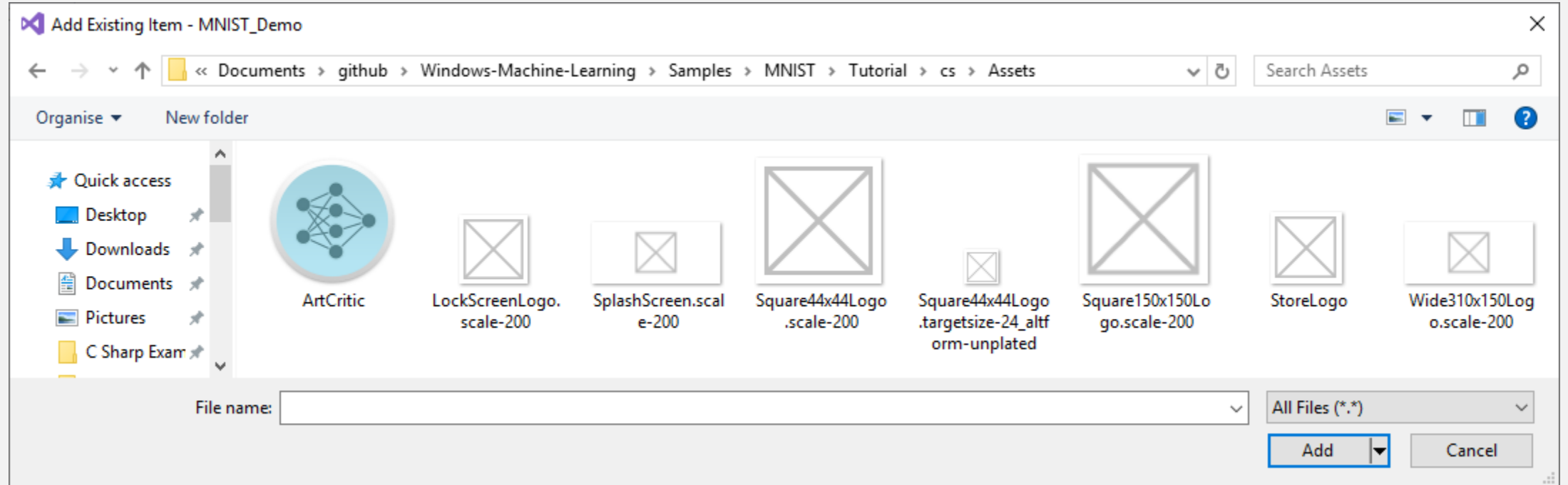
Import the model into Visual Studio



The model can now be imported into Visual Studio

Start by adding it to the Assets file for the application solution

Add the item to the project

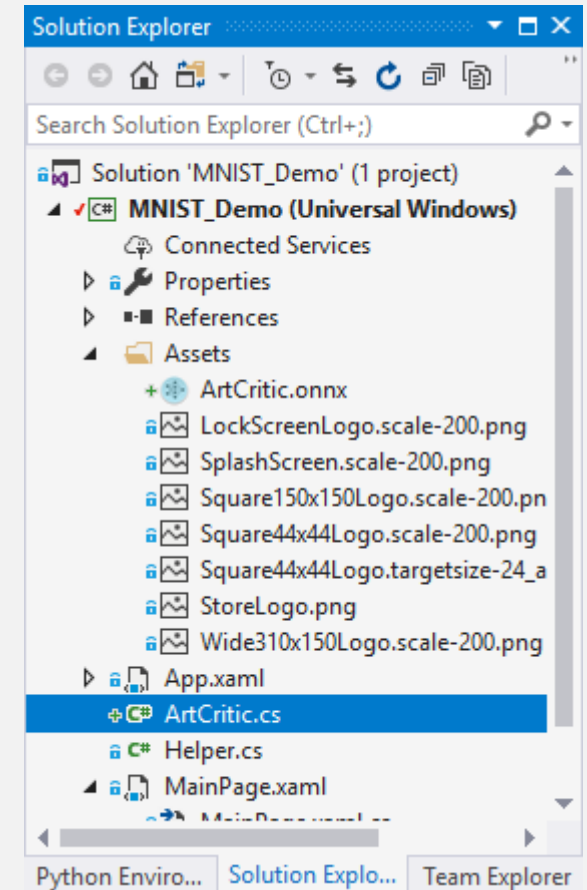


Visual Studio will import the model into the solution

The Model Interface File

When the model is imported into Visual Studio a source file is added to the project that the application can use as an interface to the model

This contains class definitions that describe the model, its input and its outputs



BIG IMPORTANT NOTE

I've had problems adding model files to brand new Universal Applications

For some reason they don't get picked up

The way round this is to start from an application like mine (which does work) and add your models to that

Using the Model

```
private ArtCriticModel ModelGen;  
private ArtCriticInput ModelInput = new ArtCriticInput();  
private ArtCriticOutput ModelOutput;
```

These variables provide the connection between the model and our application

When we want to recognise an object we need to create a ModelInput instance

This happens when the user asks the program to recognise an object

The output from the recognition is supplied in the ModelOutput instance

The program will display this result for the user

Of course your AI program could do anything it likes with the data

Using the Model

```
private async void recognizeButton_Click(object sender, RoutedEventArgs e)
{
    //Bind model input with contents from InkCanvas
    VideoFrame vf = await helper.GetHandWrittenImage(inkGrid);

    ModelInput.data = ImageFeatureValue.CreateFromVideoFrame(vf);
    // Evaluate the model
    ModelOutput = await ModelGen.EvaluateAsync(ModelInput);

    //Display the results
    numberLabel.Text = ModelOutput.classLabel.GetAsVectorView()[0];
}
```

This is the function that runs when the art critic is asked to “recognise” an image

Using the Model

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private async void recognizeButton_Click(object sender, RoutedEventArgs e)
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}
```

This code gets the image that the user has drawn

This turns out to be quite easy

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The ImageFeatureValue class is the way that a program passes images into an ML application

This code creates an instance of ImageFeatureValue from the captured video and then sets the data property of the model input to this object

Using the Model

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```

This method call is where the Model is used to analyse the input

Note that this call is performed asynchronously

It delivers an instance of the result

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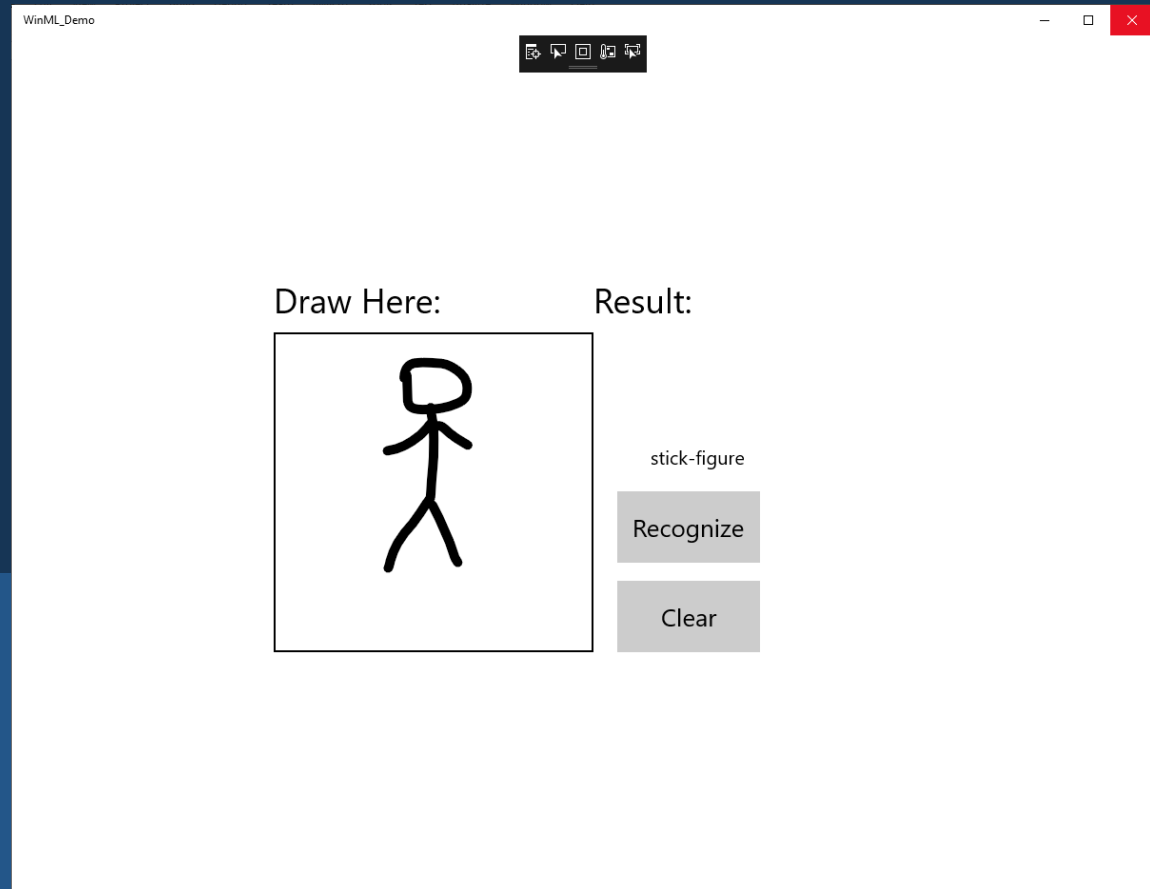
The output for this model is in the form of a TensorString

a TensorString is how the Machine Learning framework represents collections of strings

We just have to pull out one string from this, which is the name of the category recognised category

DEMO

Running the Art Critic



Summary

It is very easy to create models and use them

The model is created in the cloud but used in the device

This is “edge” ai – no need for even a network connection

The device can even be a Raspberry Pi

The ai model can run on a GPU if one is available

Visual Studio can import and use ONNX models directly and will make the wrapper classes for you

What you do with the model is entirely up to you

You can download the code and this presentation from:

<https://github.com/CrazyRobMiles/MLArt-Critic>

www.robmiles.com