



# Construyendo un detector de objetos con Azure Custom Vision .NET SDK

Luis Beltrán  
Microsoft MVP

# Luis Beltrán

- Investigador en Tomas Bata University en Zlín, República Checa.
- Docente en Tecnológico Nacional de México en Celaya, Mexico.
- Alto interés en Xamarin, Azure e Inteligencia Artificial



@darkicebeam

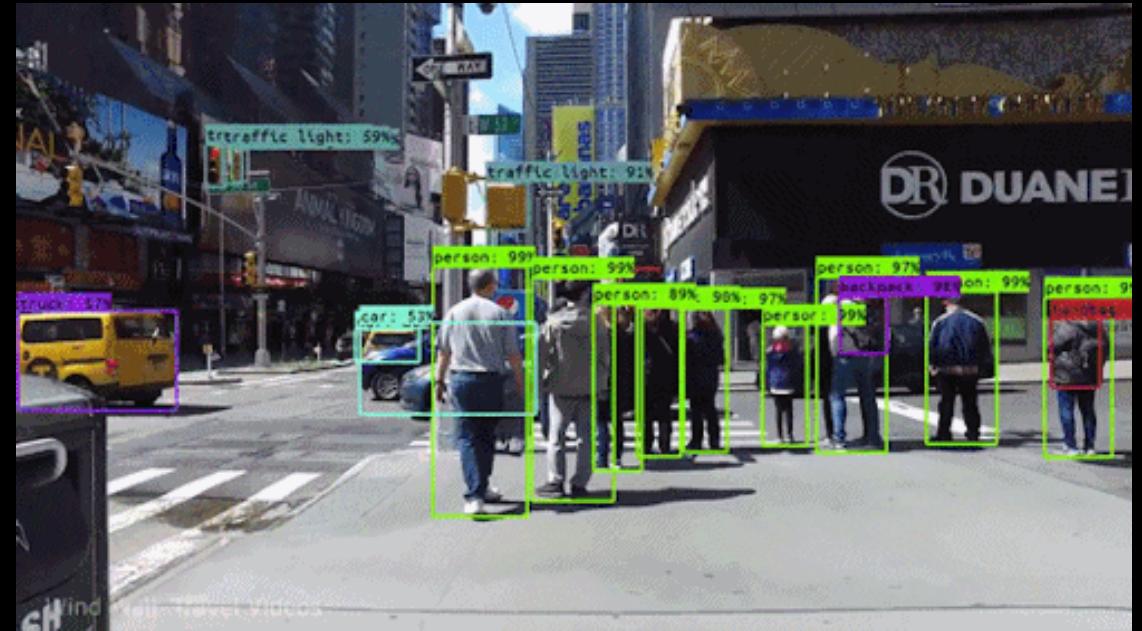


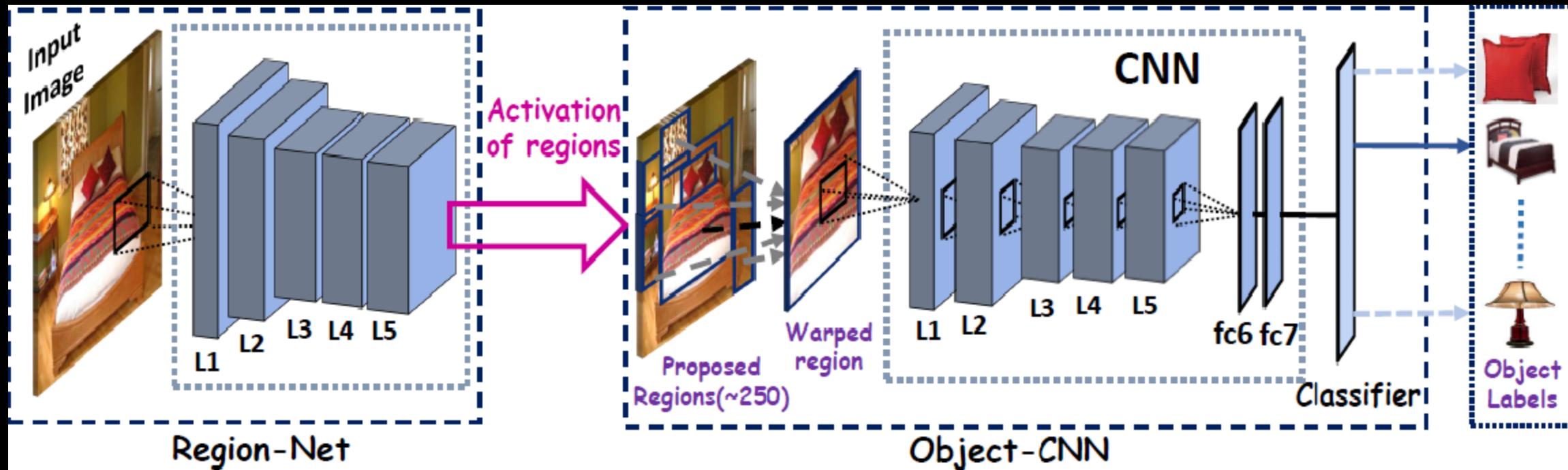
luis@luisbeltran.mx



# Detección de objetos

Es el proceso de **identificar** elementos de interés en videos e imágenes y además, de **localizarlo** dentro de la imagen/video, típicamente a través de las coordenadas de la “caja” (***bounding box***) que lo contiene.

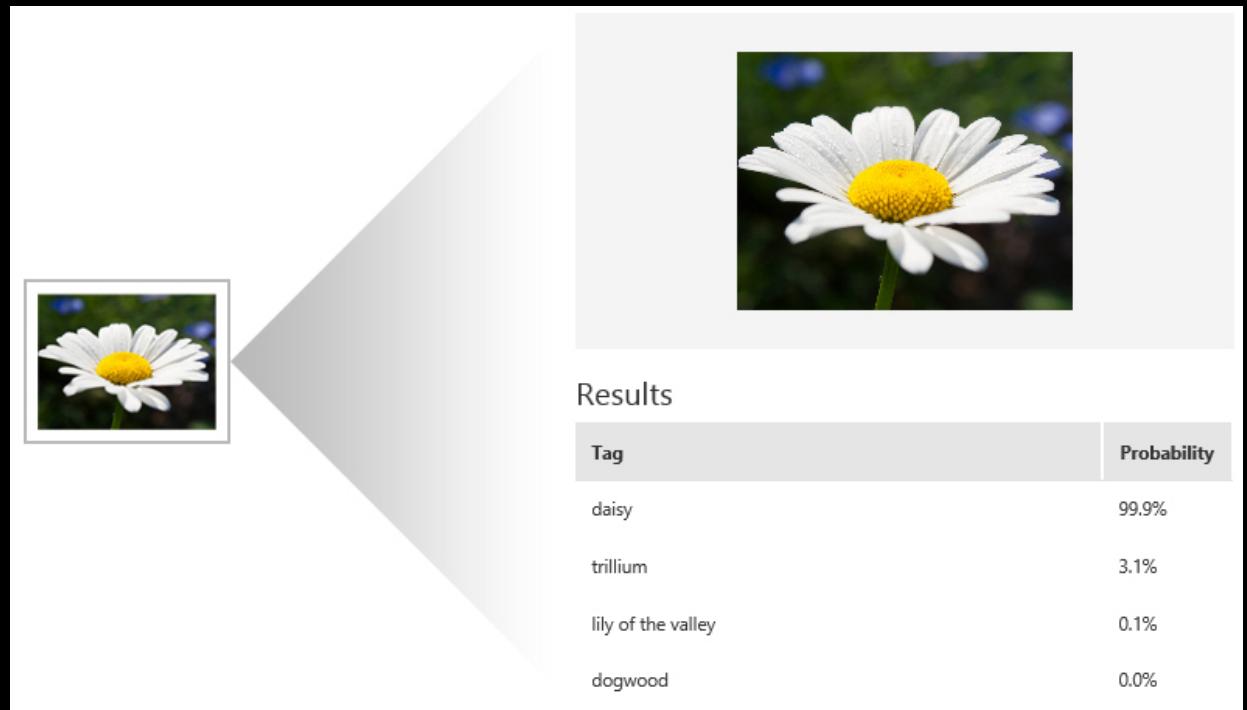




# Custom Vision Service

Reconocimiento de imágenes adaptado a las necesidades de tu caso de uso.

Un servicio y plataforma de IA para aplicar visión de computadora personalizada en tus escenarios específicos



# Construyendo un detector de objetos

## Crea un Proyecto

Elige un dominio

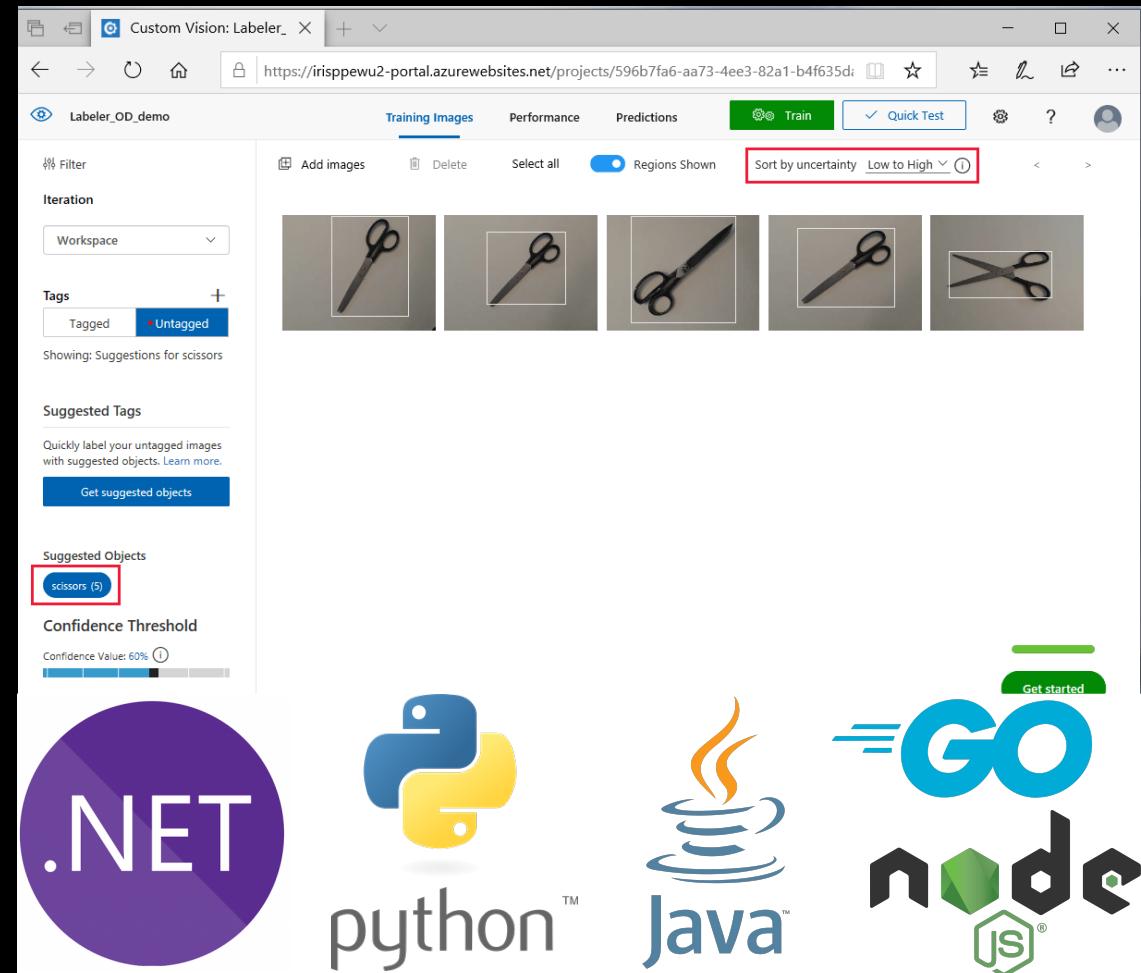
## Carga imágenes

Localiza el objeto de interés

Etiquétalo

## Entrena el detector

## Evaluá, publica y exporta el modelo



# Custom Vision .NET SDK

NuGet - Solution ➔ X Program.cs

Browse    Installed    Updates    Consolidate

custom vision       Include prerelease

Manage Packages for Solution

Package source: nuget.org

 Microsoft.Azure.CognitiveServices.Vision.CustomVision.Prediction by Microsoft, 79.5K downloads    v1.0.0  
This client library provides access to the Microsoft Cognitive Services CustomVision Prediction APIs.

 Microsoft.Azure.CognitiveServices.Vision.CustomVision.Training by Microsoft, 58.7K downloads    v1.0.0  
This client library provides access to the Microsoft Cognitive Services CustomVision Training APIs.

 Xam.Plugins.OnDeviceCustomVision by Jim Bennett, 7.04K downloads    v2.2.2  
The Azure Custom Vision service (<https://customvision.ai>) is able to create models that can be exported as CoreML, Tensorflow or ONNX models to do image classification.

 CustomVisionServiceLibrary by Marco Minerva, 772 downloads    v0.9.1  
A lightweight library that allows to easily analyze images using Custom Vision, with the ability to automatically resize the source before sending it to the service.

 Universal.Microsoft.CognitiveServices.CustomVision by Andrew Ong, 1.73K downloads    v2.4.0  
Class library for working with the Microsoft Custom Vision APIs.

 Xam.Plugins.CustomVision by Jorge Perales Diaz, 821 downloads    v4.0.2  
The Azure Custom Vision service (<https://customvision.ai>) is able to create models that can be exported as CoreML, Tensorflow or ONNX models to do image classification.

 Google.Cloud.Vision.V1 by Google LLC, 304K downloads    v2.0.0  
Recommended Google client library to access the Google Cloud Vision API, which integrates Google Vision features, including image

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 Do not show this again

 Microsoft.Azure.CognitiveServices.Vision.CustomVision.Prediction nuget.org

Versions - 0

<input checked="" type="checkbox"/> Project	Version
<input checked="" type="checkbox"/> ObjectDetectionCustomVisionApp	

Installed: not installed

Version: Latest stable 1.0.0

Options

Description

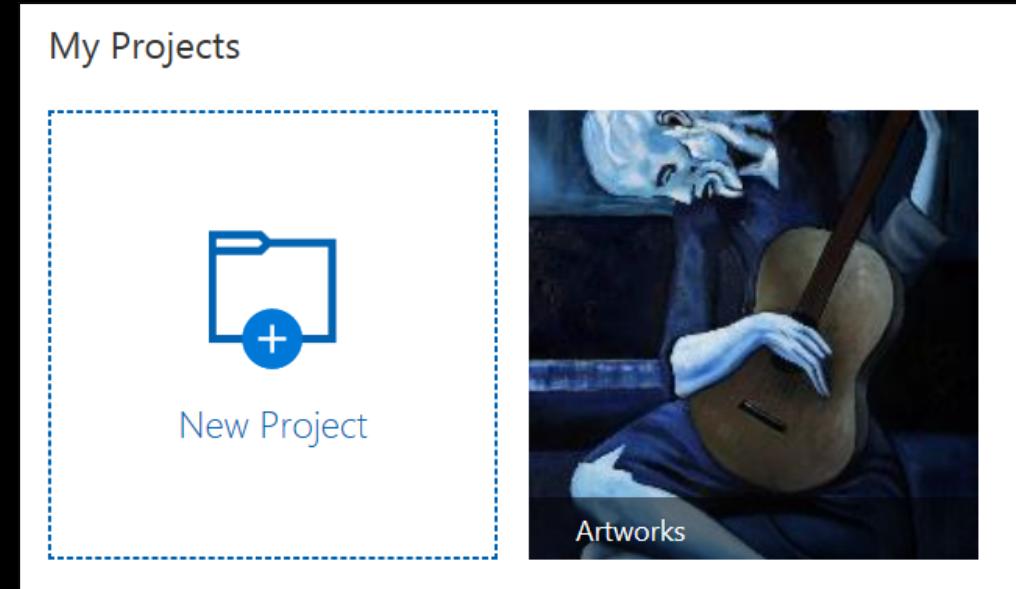
This client library provides access to the Microsoft Cognitive Services CustomVision Prediction APIs.

```
using System;
using System.IO;
using System.Linq;
using System.Text;
using System.Net.Http;
using System.Threading;
using System.Threading.Tasks;
using System.Collections.Generic;
using Microsoft.Azure.CognitiveServices.Vision.CustomVision.Training;
using Microsoft.Azure.CognitiveServices.Vision.CustomVision.Training.Models;
using Microsoft.Azure.CognitiveServices.Vision.CustomVision.Prediction;
```

# Creando un proyecto

**Custom Vision** está organizado de forma jerárquica. En el nivel más alto se encuentra el proyecto, el cual representa los datos y el modelo generado para una tarea específica.

Un **detector de objetos** es un modelo construido con el servicio de Custom Vision por medio de imágenes de entrenamiento.



Object Detector = Project

## Create new project

Name\*

Open Images Detector

Description

Open Images Custom Vision Object Detector Project

Resource

create new

OpenImagesDetectorResource [S0]

Manage Resource Permissions

Project Types 

Classification

Object Detection



## Create New Resource

Name\*

OpenImagesDetectorResource

Subscription\*

Demos Subscription

Resource Group\*

OpenImagesDetector-RG

Kind

CognitiveServices

Location

West Europe

Pricing Tier

S0

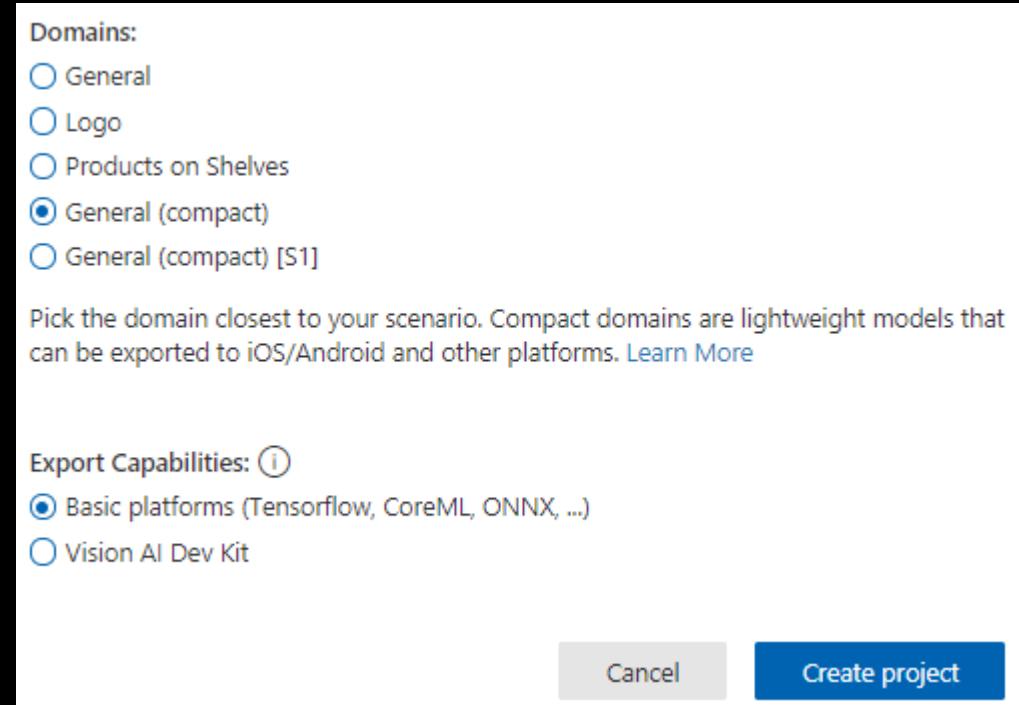
Create resource

# Eligiendo un dominio

Cuando se crea un proyecto de detección de objetos, se elige un dominio, el cual **optimiza** el detector para reconocer elementos que pertenecen a ciertos subconjuntos:

- **Logo:** logos de marcas.
- **Products on Shelves:** productos en estantes.
- **General:** para todo lo demás.

Además, existen los dominios **Compact**, que son detectores de objetos ligeros optimizados para sistemas con restricción de tiempo.



## Project Settings

### General

Project Name\*

Open Images Detector

### Resources:

OpenImagesDetectorResource

Subscription: Demos Subscription

Resource Group: OpenImagesDetector-RG

Resource Kind: All Cognitive Services

Key:

[REDACTED]

Endpoint:

https://westeurope.api.cognitive.microsoft.com/

Resource Id:

/subscriptions/[REDACTED]/resourceC

Pricing Tier: S0

1 projects created; 99 remain

Unlimited

```
// Project name must match the one in Custom Vision
var projectName = "Open Images Detector";
var publishedModelName = "OpenImagesDetectorModel";

// Replace with your own from Custom Vision project
var customVisionKey = "[REDACTED]";
var endpoint = "https://westeurope.api.cognitive.microsoft.com/";
var resourceId = "/subscriptions/[REDACTED]/resourceC/resourceGroups/[REDACTED]/providers/Microsoft.CognitiveServices/accounts/[REDACTED]";
```

```
// Training Client
var trainingApi = new CustomVisionTrainingClient()
{
    ApiKey = customVisionKey,
    Endpoint = endpoint
};

Console.WriteLine($"----- Selecting existing project: {projectName}... -----");

var projects = await trainingApi.GetProjectsAsync();
var project = projects.FirstOrDefault(x => x.Name == projectName);

if (project != null)
{
    Console.WriteLine($"\\t{projectName} found in Custom Vision workspace.");

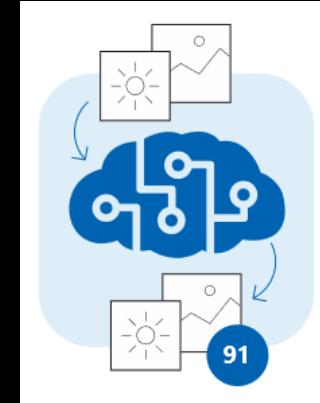
    var domainId = project.Settings.DomainId;
    var projectDomain = await trainingApi.GetDomainAsync(domainId);

    Console.WriteLine($"\\tProject domain: {projectDomain.Name}.");
    Console.WriteLine($"\\tProject type: {projectDomain.Type}.");

    WriteSeparator();
}
```

# Cargando y etiquetando imágenes

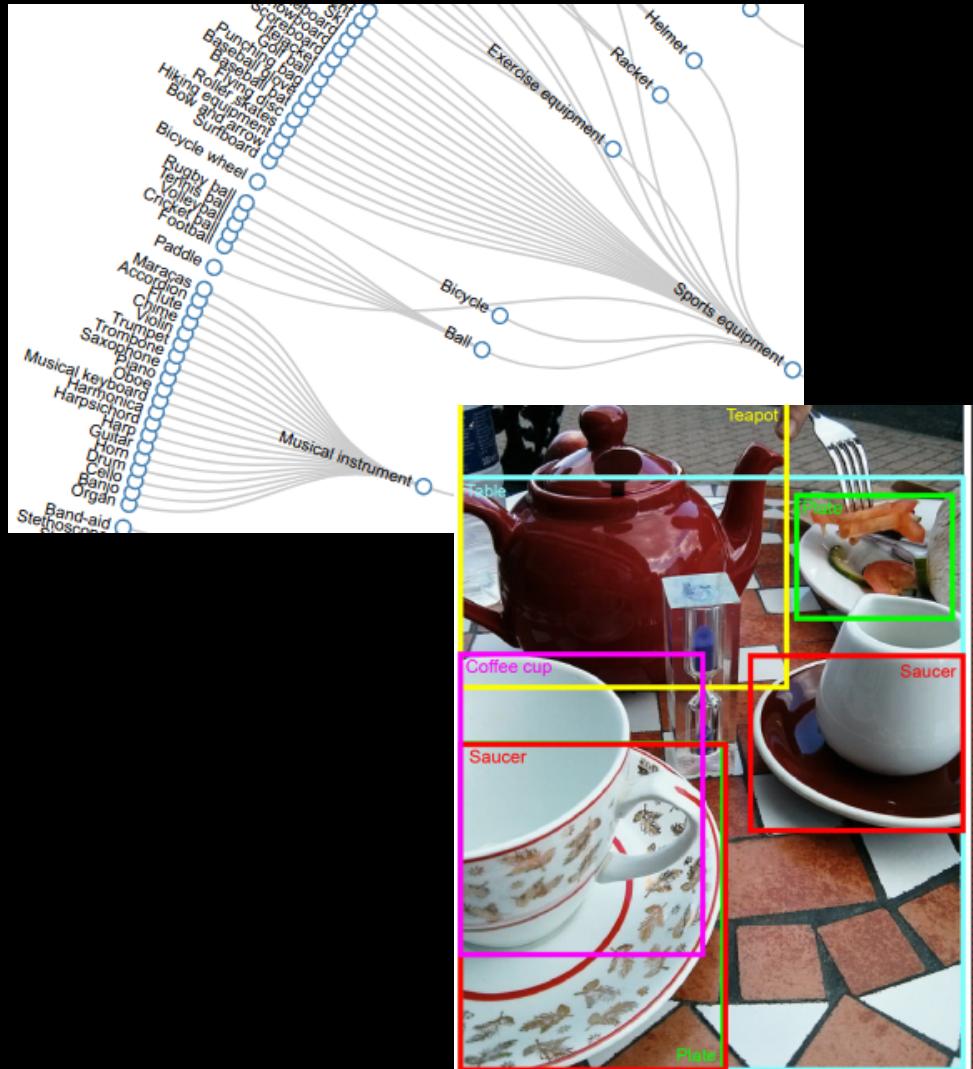
Para crear un detector de objetos con una precisión alta y confiable, Custom Vision Service requiere de varias **imágenes de entrenamiento**.



Una imagen de entrenamiento es una fotografía en la que el servicio de Custom Vision identifica de qué objeto (**tag**) se trata y su **ubicación** dentro de la imagen.



# Google Open Images Dataset V4



Open Images is a dataset of **~9M images** that have been annotated with **image-level labels**, **object bounding boxes** and **visual relationships**.

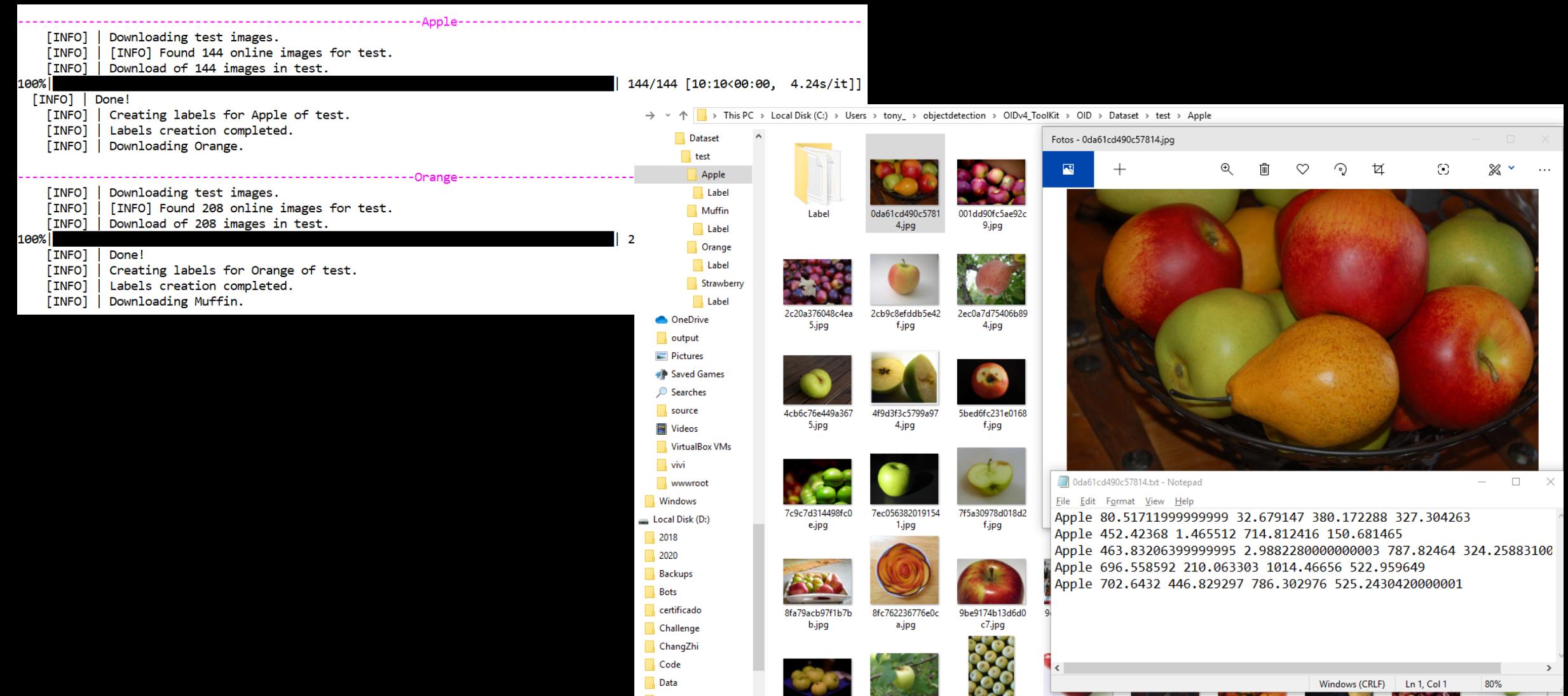
The training set of V4 contains **14.6M bounding boxes** for **600 object classes** on **1.74M images**, making it the largest existing dataset with object location annotations.

The boxes have been largely manually drawn by professional annotators to ensure accuracy and consistency.

The images are very diverse and often contain complex scenes with several objects (8.4 per image on average). This also encourages structural image annotations, such as visual relationships.

Moreover, the dataset is annotated with image-level labels spanning thousands of classes.

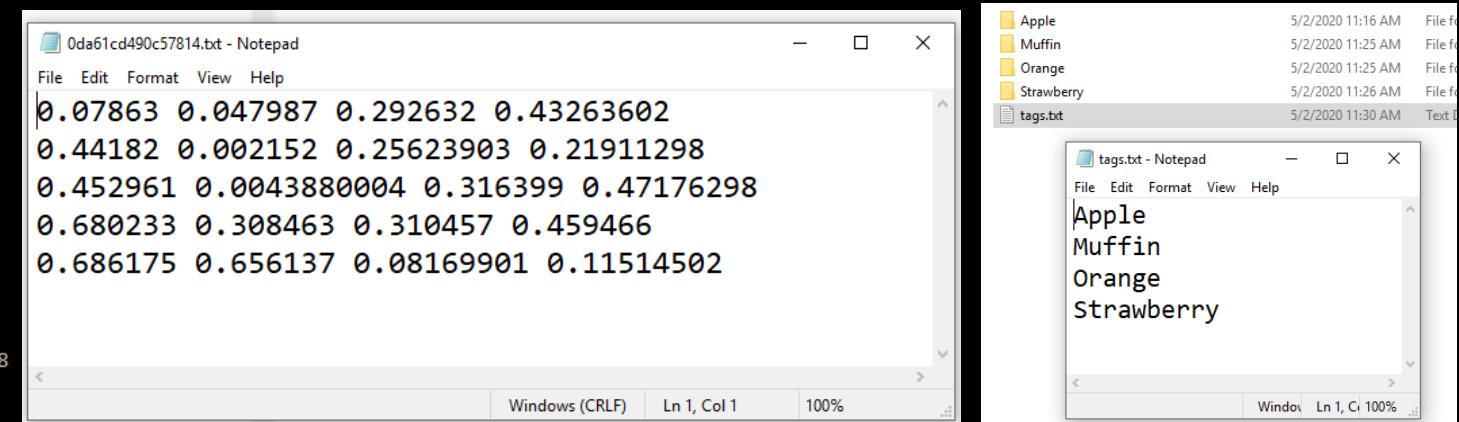
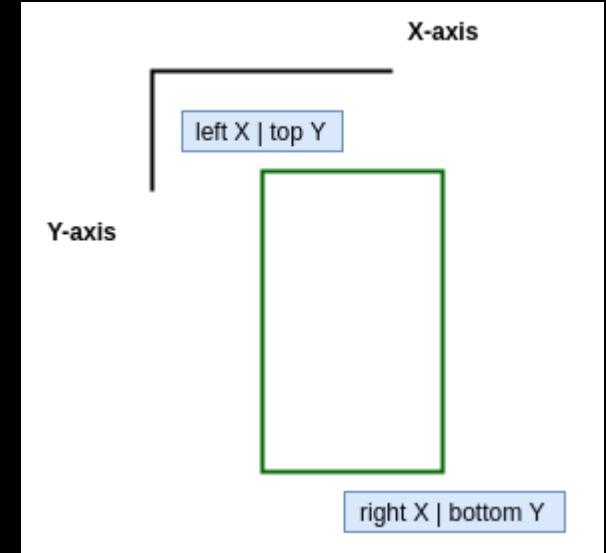
# OIDv4 Toolkit

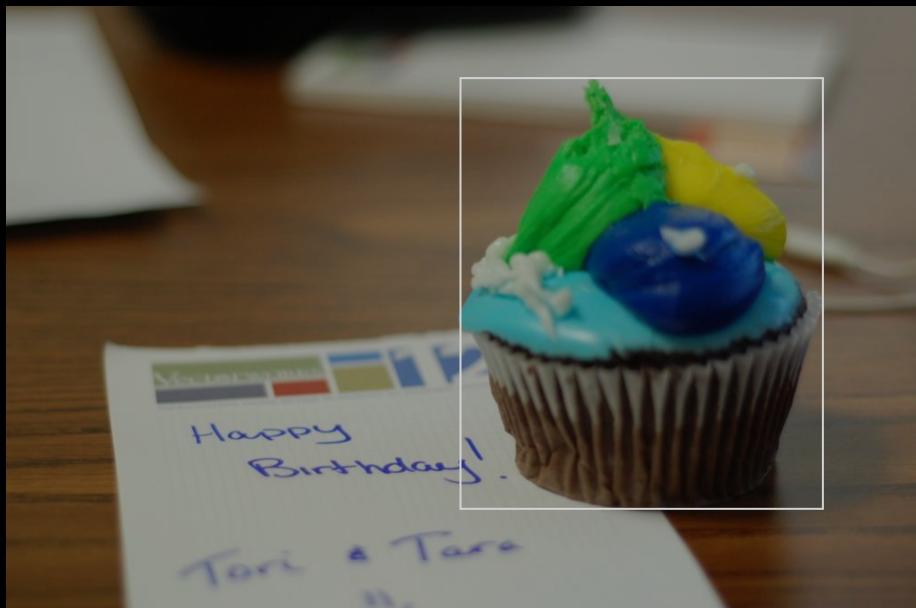
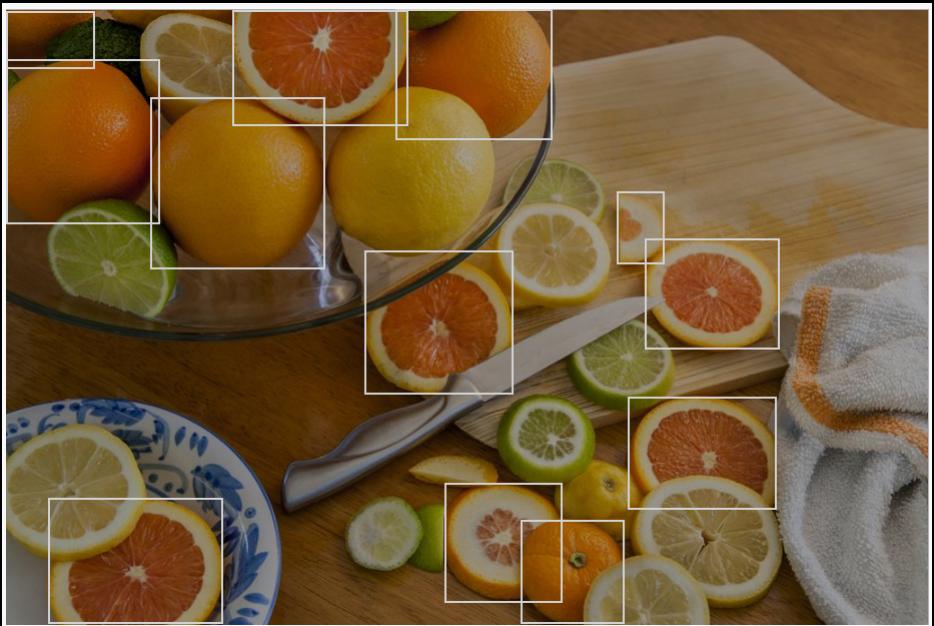
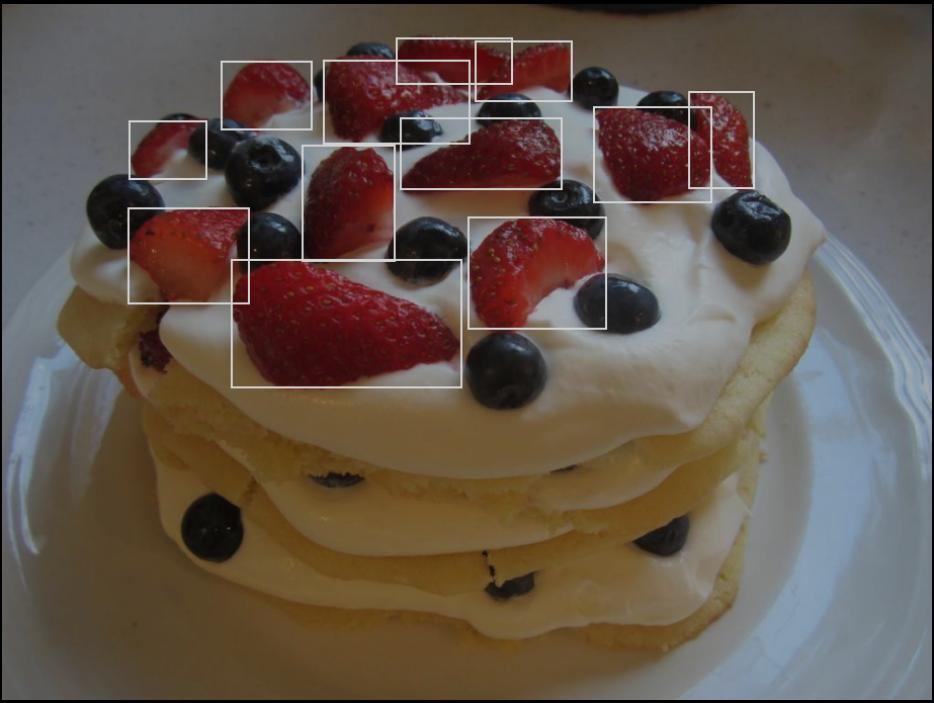


# OIDv4ToCV project

- Custom Vision requiere **coordenadas normalizadas de imagen** [0, 1] (x\_min, y\_min, width, height).
- Google Open Images Dataset v4 está **normalizado** (x\_min, x\_max, y\_min, y\_max).
- OIDv4 Toolkit genera **coordenadas desnormalizadas (pixels en la imagen)** (x\_min, y\_min, x\_max, y\_max)
- OIDv4ToCV “des-desnormaliza” las coordenadas para su uso en Custom Vision.

```
---- Apple Class-----  
Image: 001dd90fc5ae92c9.jpg  
    Normalized Data: 0 0 1 1  
Image: 08d2eed131f4b5d9.jpg  
    Normalized Data: 0 0 0.220249 0.326184  
    Normalized Data: 0.001591 0.337568 0.177223 0.339766  
Image: 096ca56a5c51f6d6.jpg  
    Normalized Data: 0 0.00095699995 1 0.999043  
Image: 0da61cd490c57814.jpg  
    Normalized Data: 0.07863 0.047987 0.292632 0.43263602  
    Normalized Data: 0.44182 0.002152 0.25623903 0.21911298  
    Normalized Data: 0.452961 0.0043880004 0.316399 0.47176298  
    Normalized Data: 0.680233 0.308463 0.310457 0.459466  
    Normalized Data: 0.686175 0.656137 0.08169901 0.11514502  
Normalized Data: 0.686175 0.656137 0.08169901 0.11514502  
----
```



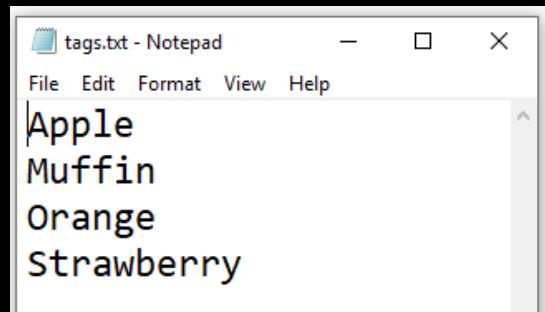


```
// Retrieve the tags that already exist in the project
Console.WriteLine("----- Retrieving tags... -----");

var modelTags = await trainingApi.GetTagsAsync(project.Id);

var tags = new List<Tag>();
var onlineImages = 0;

// Obtain tags from our dataset
var tagsFile = Path.Combine(trainingFolder, "tags.txt");
var imageLabels = await File.ReadAllLinesAsync(tagsFile);
```



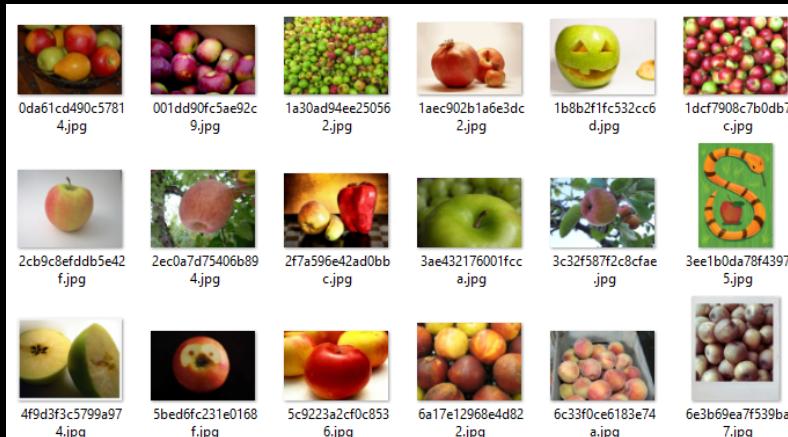
```
foreach (var label in imageLabels)
{
    // Check if the label already exists
    var tag = modelTags.FirstOrDefault(x => x.Name == label);

    if (tag == null)
    {
        // If not, create it
        tag = await trainingApi.CreateTagAsync(project.Id, label);

        Console.WriteLine($"\\tTag {tag.Name} was created.");
    }
    else
    {
        // If so, just count images with this tag
        onlineImages += tag.ImageCount;
        Console.WriteLine($"\\tTag {label} was NOT created (it already exists)");
    }

    tags.Add(tag);
}
```

- Apple
- Muffin
- Orange
- Strawberry



```
0da61cd490c57814.txt - Notepad
File Edit Format View Help
0.07863 0.047987 0.292632 0.43263602
0.44182 0.002152 0.25623903 0.21911298
0.452961 0.0043880004 0.316399 0.47176298
0.680233 0.308463 0.310457 0.459466
0.686175 0.656137 0.08169901 0.11514502
```

```
Console.WriteLine("----- Accessing images -----");
var imageFileEntries = new List<ImageFileCreateEntry>();

foreach (var label in imageLabels)
{
    var tagFolder = Path.Combine(trainingFolder, label);
    var tagImages = Directory.GetFiles(tagFolder);
    var tagLabels = Path.Combine(tagFolder, "normalizedLabel");

    foreach (var image in tagImages)
    {
        var imageName = Path.GetFileNameWithoutExtension(image);
        var imageFile = Path.GetFileName(image);
        var imageLabelFile = $"{imageName}.txt";
        var tagCV = tags.Single(x => x.Name == label);

        var imageBoundingBoxes = await File.ReadAllLinesAsync(Path.Combine(tagLabels, imageLabelFile));
        var imageRegions = new List<Region>();

        foreach (var bbox in imageBoundingBoxes)
        {
            var normalizedData = bbox.Split();
            var left = float.Parse(normalizedData[0]);
            var top = float.Parse(normalizedData[1]);
            var width = float.Parse(normalizedData[2]);
            var height = float.Parse(normalizedData[3]);

            imageRegions.Add(new Region(tagCV.Id, left, top, width, height));
        }

        Console.WriteLine($"\\tAdding image {imageName} with its regions.");
        imageFileEntries.Add(new ImageFileCreateEntry(
            imageFile, await File.ReadAllBytesAsync(image), null, imageRegions));
    }
}
```

```
var batchPageSize = 64;
var numBatches = imageFileEntries.Count / batchPageSize;
var sizeLastBatch = imageFileEntries.Count % batchPageSize;

for (int batch = 0; batch < numBatches; batch++)
{
    Console.WriteLine($"\\tUploading images batch #{batch}.");

    var entries = imageFileEntries.Skip(batch * batchPageSize).Take(batchPageSize).ToList();

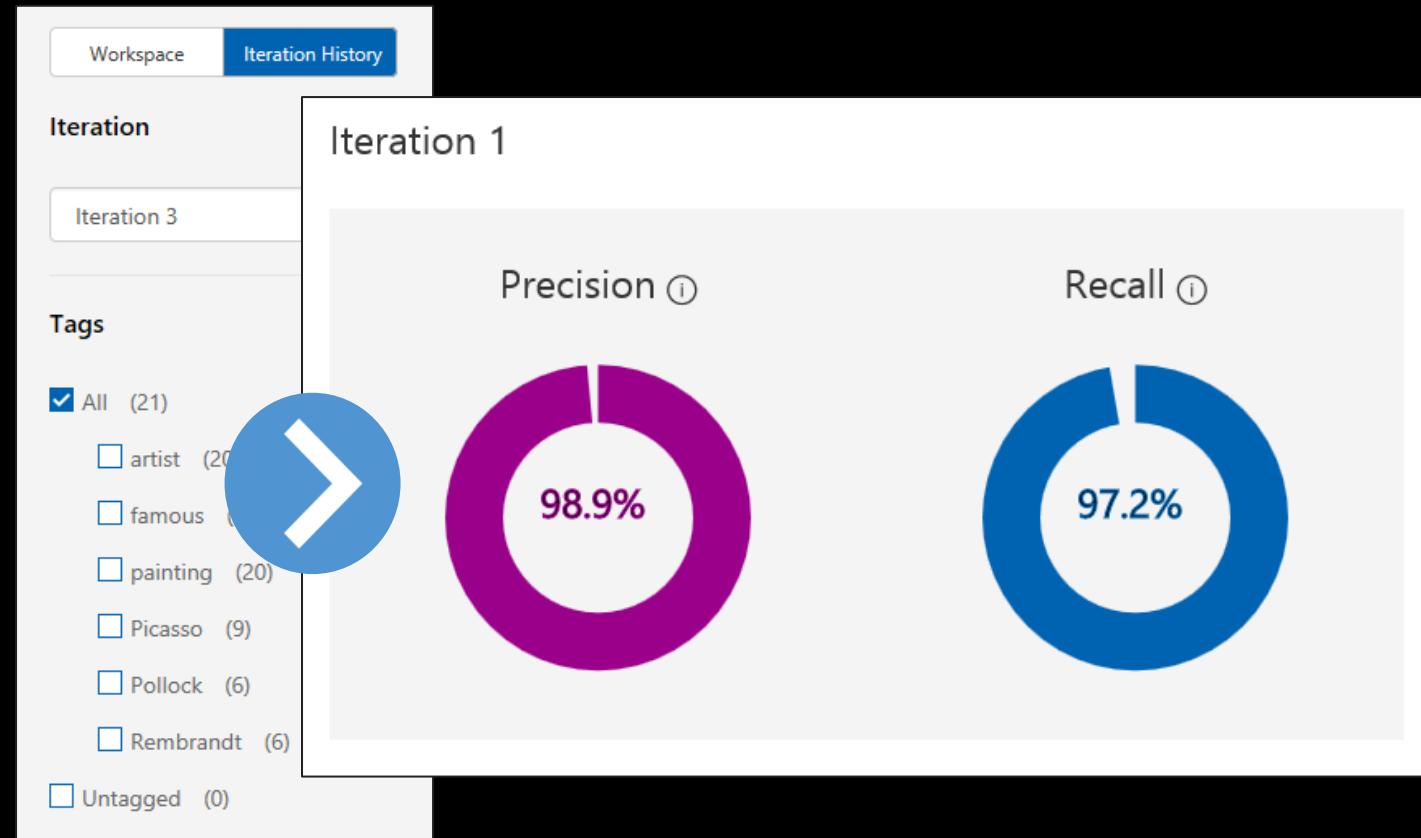
    await trainingApi.CreateImagesFromFilesAsync(
        project.Id, new ImageFileCreateBatch(entries));
}

if (sizeLastBatch > 0)
{
    Console.WriteLine($"\\tUploading last batch.");
    var lastEntries = imageFileEntries.Skip(numBatches * batchPageSize).Take(sizeLastBatch).ToList();

    await trainingApi.CreateImagesFromFilesAsync(project.Id,
        new ImageFileCreateBatch(lastEntries));
}
```

# Entrenando el detector de objetos

Cada vez que se (re)entrena el detector, se está creando una iteración del modelo.



```
// Now there are images with tags start training the project
Console.WriteLine("----- Starting the Training process... -----");

iteration = await trainingApi.TrainProjectAsync(project.Id);

// The returned iteration will be in progress, and can be queried periodically to see when it has completed
while (iteration.Status == "Training")
{
    Thread.Sleep(1000);
    Console.WriteLine($"\\tIteration '{iteration.Name}' status: {iteration.Status}");

    // Re-query the iteration to get its updated status
    iteration = await trainingApi.GetIterationAsync(project.Id, iteration.Id);
}

Console.WriteLine($"\\tIteration '{iteration.Name}' status: {iteration.Status}");
```

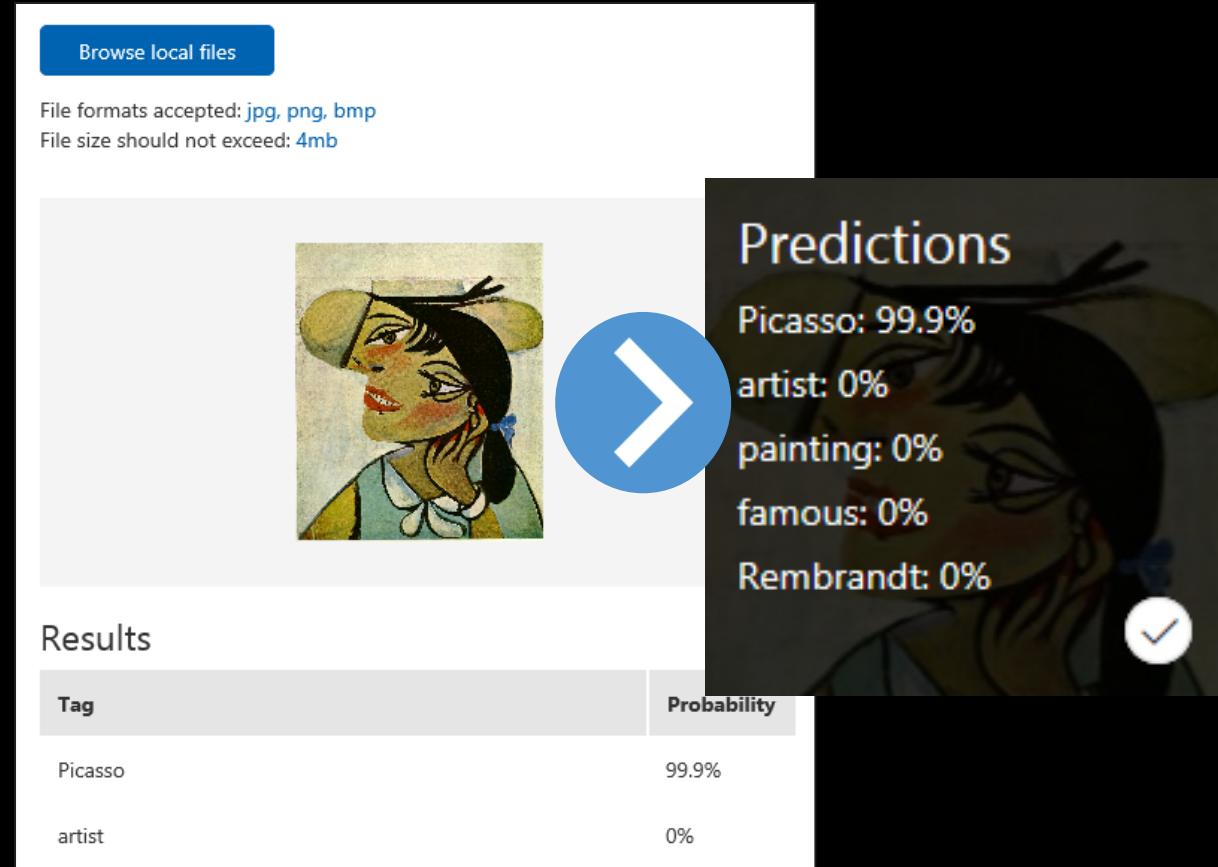
# Evaluando y probando el modelo

Después de entrenar el modelo, éste puede **probarse** rápidamente con nuevas imágenes.

**La prueba utiliza la última iteración (la más reciente).**

Si el modelo proporciona resultados precisos, puede ser publicado, haciendo que el detector:

- Pueda **accederse** desde un **endpoint HTTP** (o a través del **SDK**) para detección de objetos online, a través de Internet.
- Pueda **exportarse** a una plataforma para una detección de objetos sin requerir Internet (en modo offline).



```
// The iteration is now trained. Publish it to the prediction endpoint.  
Console.WriteLine($"----- Starting the Publication process. -----");  
  
await trainingApi.PublishIterationAsync(  
    project.Id, iteration.Id, publishedModelName, resourceId);  
  
Console.WriteLine($"\\tIteration '{iteration.Name}' published.");
```

Open Images Detector

Training Images Performance Predictions Train Quick Test ?

Iterations Unpublish Prediction URL Delete Export

Probability Threshold: 50% (i)

Overlap Threshold: 30% (i)

Iteration 1 PUBLISHED

Trained : 1 hours ago with General (compact) domain

Iteration 1

Finished training on 5/2/2020, 5:23:51 PM using General (compact) domain

Iteration id:

Published a

How to use the Prediction API

If you have an image URL:

```
https://westeurope.api.cognitive.microsoft.com/customvision/v3.0/Prediction/2fa091  
Set Prediction-Key Header to: [REDACTED]  
Set Content-Type Header to: application/json  
Set Body to: {"Url": "https://example.com/image.png"}
```

If you have an image file:

```
https://westeurope.api.cognitive.microsoft.com/customvision/v3.0/Prediction/2fa091  
Set Prediction-Key Header to: [REDACTED]  
Set Content-Type Header to: application/octet-stream  
Set Body to: <image file>
```

Got it!

Performance

Tag	Precision	Recall	A.P.	Image count
Muffin	62.8%	42.9%	53.4%	128
Strawberry	60.0%	19.9%	33.3%	225
Apple	52.2%	15.6%	25.7%	144

Get started

```
// Prediction Client
var predictionClient = new CustomVisionPredictionClient()
{
    ApiKey = customVisionKey,
    Endpoint = endpoint
};

foreach (var image in testImages)
{
    var imageName = Path.GetFileName(image);

    using (var stream = new MemoryStream(File.ReadAllBytes(image)))
    {
        Console.WriteLine($"\\tImage: {imageName}");

        var result = await predictionClient.DetectImageAsync(
            project.Id, publishedModelName, stream);

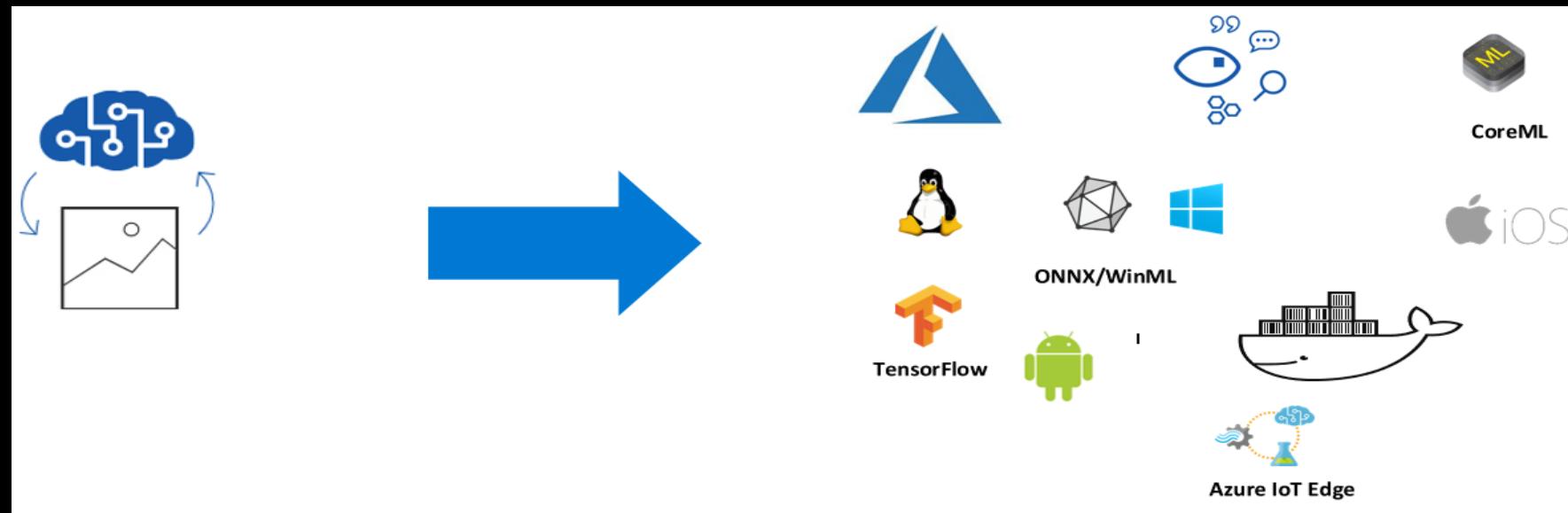
        // Loop over each prediction and write out the results
        foreach (var prediction in result.Predictions.OfType<ImagePrediction>().OrderByDescending(x => x.Probability))
        {
            Console.WriteLine($"\\t\\tFor Tag '{prediction.TagName}': {prediction.Probability:P3} " +
                $"[{prediction.BoundingBox.Left}, {prediction.BoundingBox.Top}, " +
                $" {prediction.BoundingBox.Width}, {prediction.BoundingBox.Height}]");
        }

        WriteSeparator();
        Console.WriteLine("Press any key for next image...");
        Console.ReadKey();
    }
}
```

# Ejecuta el modelo donde deseas

Ejecuta tus modelos donde los necesites, de acuerdo a tus escenarios y requerimientos únicos.

Exporta tus modelos entrenados a dispositivos o contenedores para escenarios de baja latencia.



- Domains:
- General
  - Logo
  - Products on Shelves
  - General (compact)
  - General (compact) [S1]

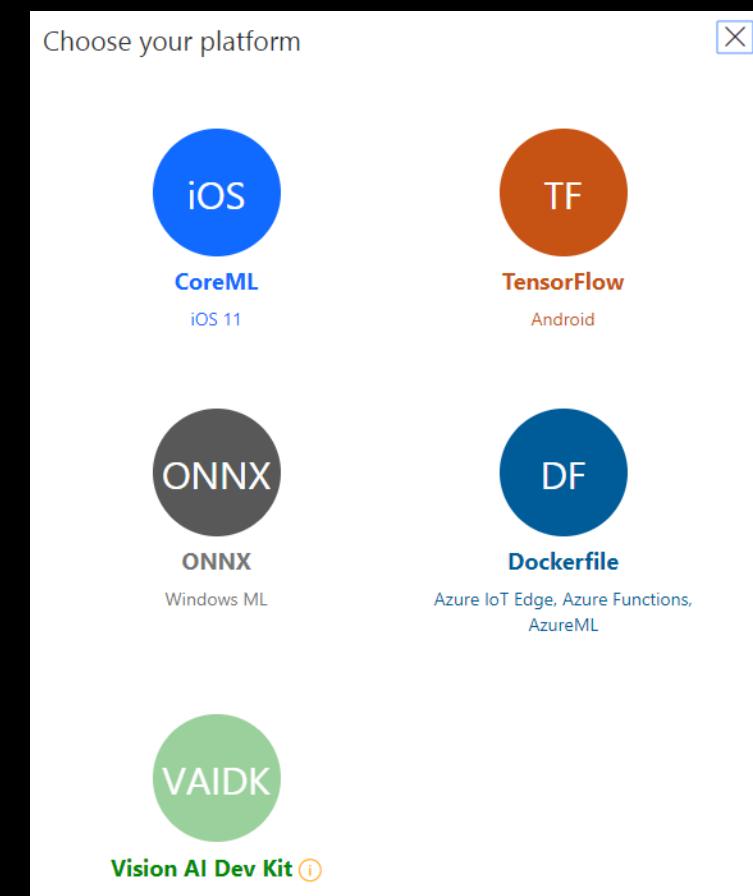
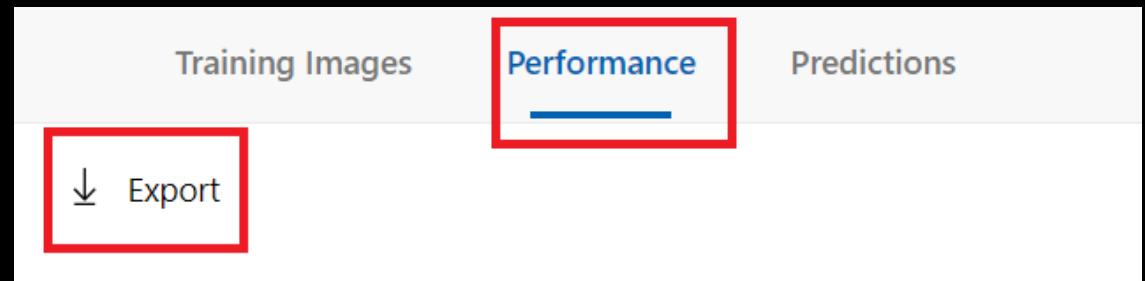
Pick the domain closest to your scenario. Compact domains are lightweight models that can be exported to iOS/Android and other platforms. [Learn More](#)

Export Capabilities: ⓘ

- Basic platforms (Tensorflow, CoreML, ONNX, ...)
- Vision AI Dev Kit

[Cancel](#)

[Create project](#)



```
platform = "TensorFlow";
extension = "zip";
```

```
do
{
    var exports = await trainingApi.GetExportsAsync(project.Id, iteration.Id);

    export = exports.FirstOrDefault(x => x.Platform == platform);

    if (export == null)
        export = await trainingApi.ExportIterationAsync(project.Id, iteration.Id, platform);

    Thread.Sleep(1000);
    Console.WriteLine($"\\tStatus: {export.Status}");
} while (export.Status == "Exporting");
```

```
if (export.Status == ExportStatus.Done)
{
    Console.WriteLine($"\\tDownloading {platform} model");
    var filePath = Path.Combine(Environment.CurrentDirectory, $"{publishedModelName}_{platform}.{extension}");

    using (var httpClient = new HttpClient())
    {
        using (var stream = await httpClient.GetStreamAsync(export.DownloadUri))
        {
            using (var file = new FileStream(filePath, FileMode.Create))
            {
                await stream.CopyToAsync(file);
                Console.WriteLine($"\\tModel successfully exported. You can find it here:\\n\\t{filePath}.");
                WriteSeparator();
            }
        }
    }
}
```

# Demo: Creando un detector de objetos con Custom Vision .NET SDK

## Información del proyecto

```
----- Selecting existing project: Open Images Detector... -----  
Open Images Detector found in Custom Vision workspace.  
Project domain: General (compact).  
Project type: ObjectDetection.
```

## Añadiendo etiquetas al modelo

```
----- Retrieving tags... -----  
Tag Apple was created.  
Tag Muffin was created.  
Tag Orange was created.  
Tag Strawberry was created.
```

## Cargando imágenes de entrenamiento

```
----- Accessing images -----  
Adding image 001dd90fc5ae92c9 with its regions.  
Adding image 08d2eed131f4b5d9 with its regions.  
Adding image 096ca56a5c51f6d6 with its regions.  
Adding image 0da61cd490c57814 with its regions.  
Adding image 10628bb9eeafafbb with its regions.  
Adding image 1124df73ba9c7557 with its regions.  
Adding image 123c5bfc13f1c62d with its regions.  
Adding image 12fb96c87a65c542 with its regions.  
Adding image 13628bbfec1fe2c3 with its regions.
```

```
Uploading images batch #0.  
Uploading images batch #1.  
Uploading images batch #2.  
Uploading images batch #3.  
Uploading images batch #4.  
Uploading images batch #5.  
Uploading images batch #6.  
Uploading images batch #7.  
Uploading images batch #8.  
Uploading images batch #9.  
Uploading images batch #10.  
Uploading last batch.
```

## Entrenando el modelo

```
----- Starting the Training process... -----  
Iteration 'Iteration 1' status: Training  
  
Iteration 'Iteration 1' status: Training  
Iteration 'Iteration 1' status: Training  
Iteration 'Iteration 1' status: Training  
Iteration 'Iteration 1' status: Training  
Iteration 'Iteration 1' status: Training  
Iteration 'Iteration 1' status: Training  
Iteration 'Iteration 1' status: Completed
```

## Publicando el modelo

```
----- Starting the Publication process. -----  
Iteration 'Iteration 1' published.  
-----
```

## Evaluando el modelo con imágenes de prueba



```
----- Making predictions -----  
Image: Test01.jpg  
For Tag 'Muffin': 97.467% [0.12562418, 0.266169071, 0.6825628, 0.6615833]  
For Tag 'Muffin': 4.869% [0.331231415, 0.61405015, 0.6362891, 0.385939837]  
For Tag 'Muffin': 3.135% [0.174345419, 0.409970373, 0.458655715, 0.218689173]  
For Tag 'Muffin': 2.991% [0.008214988, 0.03092061, 0.181516379, 0.257676244]  
For Tag 'Muffin': 2.613% [0.12349081, 0.6451644, 0.681633651, 0.354825616]  
For Tag 'Muffin': 2.177% [0.3288841, 0.47192055, 0.164117157, 0.100035369]  
For Tag 'Muffin': 1.830% [0.1604278, 0, 0.4830678, 0.0725843]  
For Tag 'Muffin': 1.697% [0, 0.005163975, 0.2836524, 0.197787941]  
For Tag 'Muffin': 1.292% [0.452365875, 0.336768657, 0.500448346, 0.5399126]  
For Tag 'Muffin': 1.246% [0.637812257, 0, 0.36217773, 0.06886031]  
For Tag 'Muffin': 1.100% [0.0481789559, 0.474761367, 0.190935969, 0.122284174]  
For Tag 'Strawberry': 1.082% [0, 0.0141699323, 0.07569957, 0.0172690973]
```

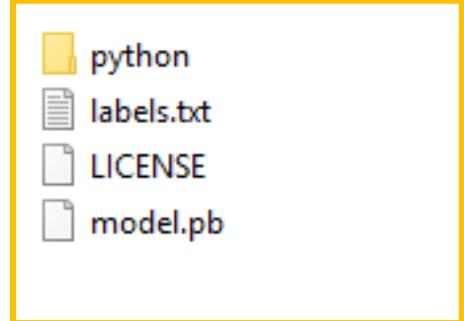
-----

```
Image: Test02.jpg  
For Tag 'Apple': 92.286% [0.66393584, 0.322840065, 0.3145402, 0.551750064]  
For Tag 'Apple': 82.698% [0.215984285, 0.409043372, 0.327929437, 0.421557963]  
For Tag 'Apple': 82.100% [0.43457824, 0.07026026, 0.3172269, 0.502035737]  
For Tag 'Apple': 62.339% [0.0130720735, 0.09492104, 0.176995143, 0.456798434]  
For Tag 'Apple': 10.023% [0.202900678, 0.138997942, 0.542853832, 0.8015692]  
For Tag 'Apple': 4.737% [0.9709588, 0.455433547, 0.0290311575, 0.362548053]  
For Tag 'Apple': 3.711% [0, 0, 0.2653275, 0.656832933]  
For Tag 'Apple': 2.553% [0.352338046, 0, 0.593947768, 0.890323639]  
For Tag 'Apple': 1.455% [0.9459655, 0.293046474, 0.0540244579, 0.49333632]  
For Tag 'Apple': 1.390% [0.120855212, 0.227188244, 0.06773448, 0.303626239]
```

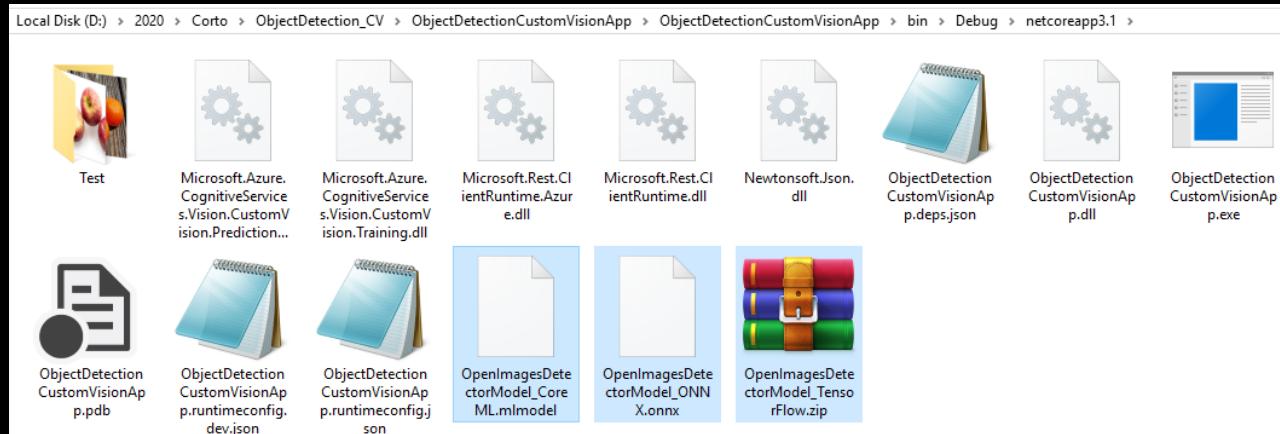
-----

## Exportando el modelo a varias plataformas

```
---- Do you want to export the model? (Y/N) ----  
Y Options:  
 1) TensorFlow  
 2) CoreML  
 3) Other platform  
 E) End program  
1-----  
  Exporting to TensorFlow...  
  Status: Exporting  
  Status: Done  
Status: Done  
  Downloading TensorFlow model  
  Model successfully exported. You can find it here:  
  D:\2020\Corito\ObjectDetection_CV\ObjectDetectionCustomVisionApp\ObjectDetectionCustomVisionApp\bin\Debug\netcoreapp3.1\OpenImagesDetectorModel_TensorFlow.zip.
```

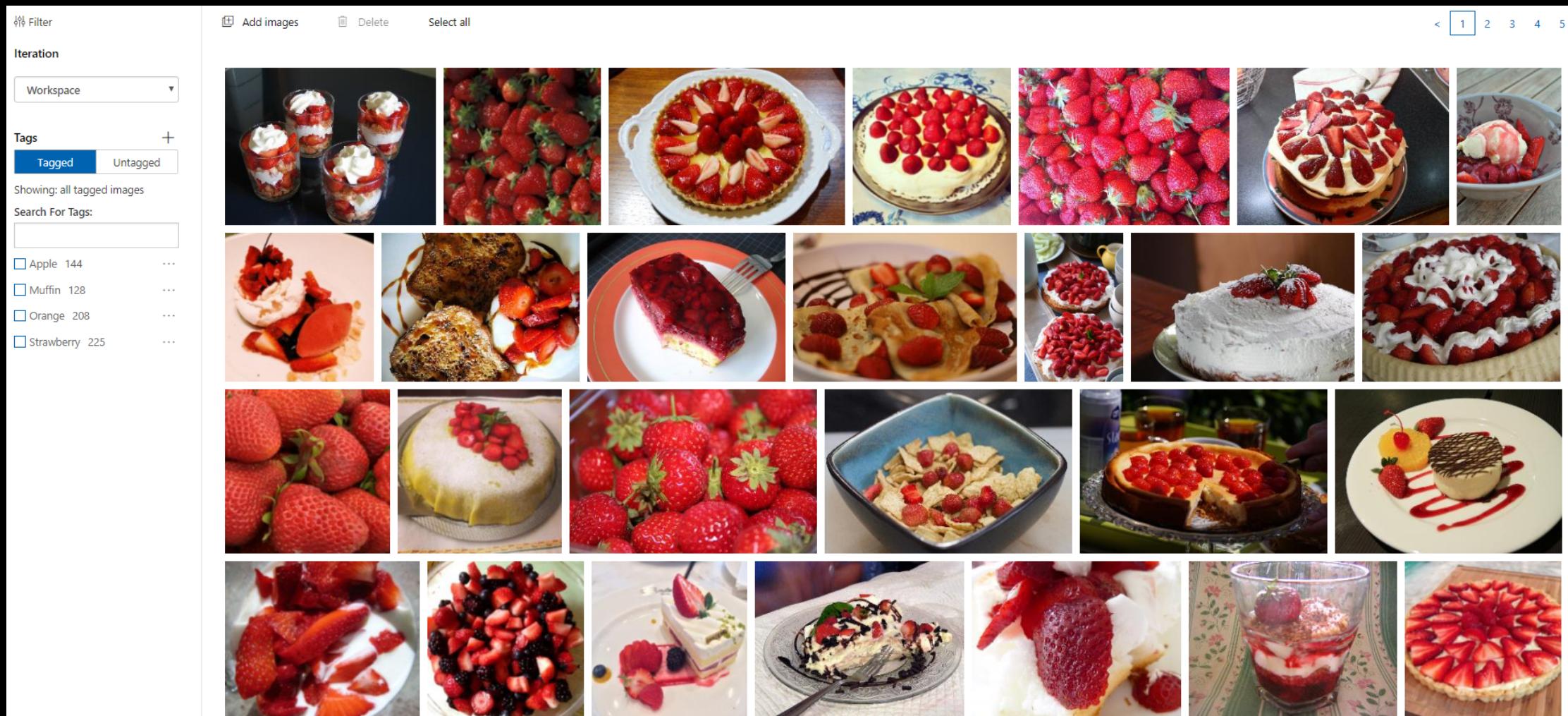


```
Options:  
 1) TensorFlow  
 2) CoreML  
 3) Other platform  
 E) End program  
2-----  
  Exporting to CoreML...  
  Status: Exporting  
  Status: Done  
Status: Done  
  Downloading CoreML model  
  Model successfully exported. You can find it here:  
  D:\2020\Corito\ObjectDetection_CV\ObjectDetectionCustomVisionApp\ObjectDetectionCustomVisionApp\bin\Debug\netcoreapp3.1\OpenImagesDetectorModel_CoreML.mlmodel.
```



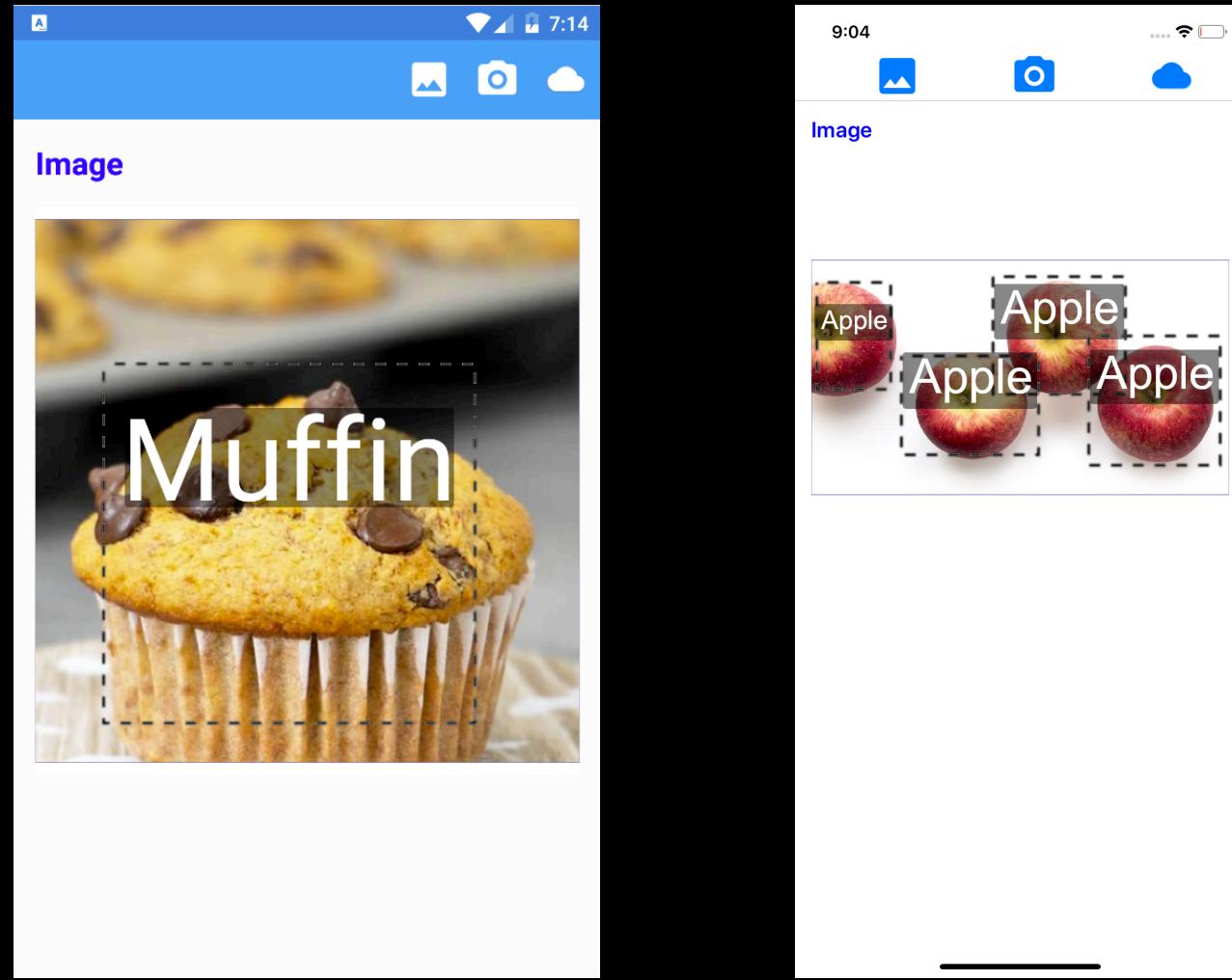
```
---- Do you want to export the model? (Y/N) ----  
y Options:  
 1) TensorFlow  
 2) CoreML  
 3) Other platform  
 E) End program  
3-----  
  ONNX  
  Type the platform name  
  Now type the file extension for the ONNX exported model.  
onnx-----  
  Exporting to ONNX...  
  Status: Done  
Status: Done  
  Downloading ONNX model  
  Model successfully exported. You can find it here:  
  D:\2020\Corito\ObjectDetection_CV\ObjectDetectionCustomVisionApp\ObjectDetectionCustomVisionApp\bin\Debug\netcoreapp3.1\OpenImagesDetectorModel_ONNX.onnx.
```

## Estado del proyecto en el portal de Custom Vision después de crear el modelo entrenado



Proyecto open-source disponible en GitHub: <https://github.com/icebeam7/ObjectDetectionCustomVisionApp>

# Demo: Detección de objetos en apps móviles con Custom Vision y Xamarin



Proyecto open-source disponible en GitHub: <https://github.com/icebeam7/ObjectDetectionMobileApp>

The screenshot shows the Visual Studio Enterprise 2019 for Mac interface. On the left is the Solution Explorer with project files for 'ObjectDetectionMobileApp' including 'CustomVisionAzureService.cs'. The main window displays the 'CustomVisionAzureService.cs' code:

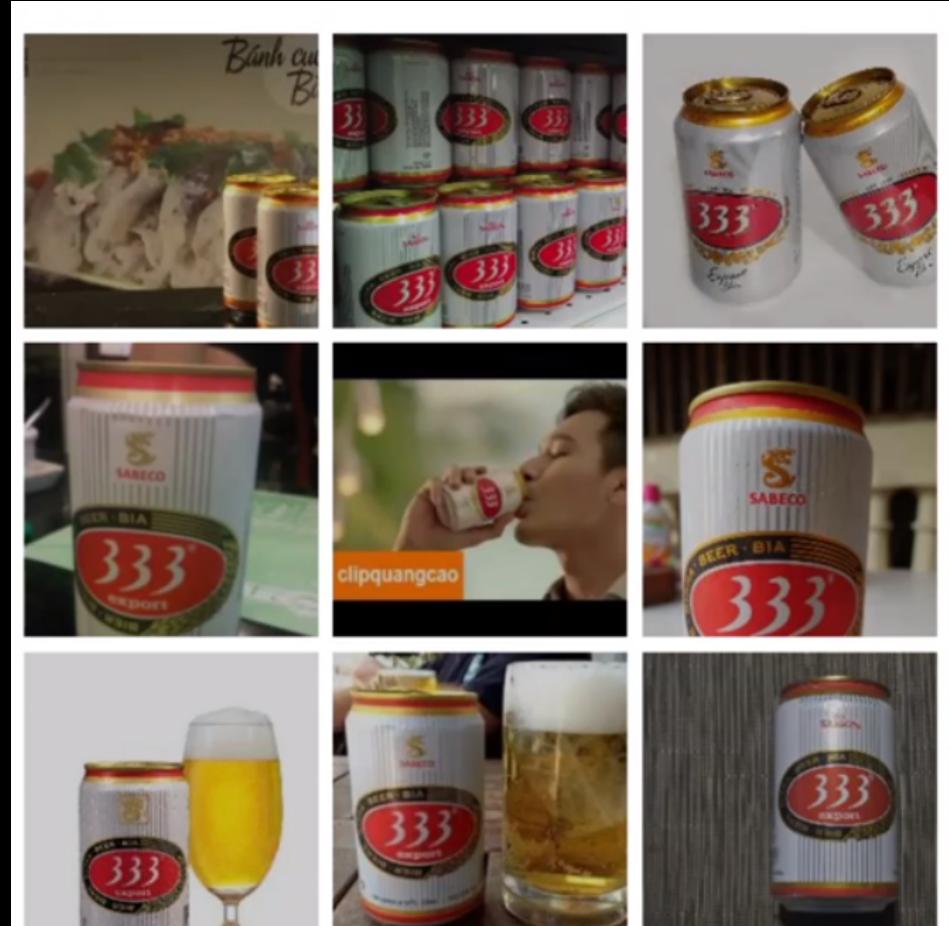
```
28     return client;
29 }
30 }
31 public async static Task<List<Prediction>> DetectObjects(MediaFile picture)
32 {
33     try
34     {
35         ServicePointManager.SecurityProtocol = SecurityProtocol.Tls12 | SecurityProtocol.Tls11 | SecurityProtocol.Tls;
36
37         var stream = new MemoryStream();
38         await picture.GetStreamWithImageRotatedForExternalStorageAsync(stream);
39
40         using (var content = new ByteArrayContent(stream.ToArray()))
41         {
42             content.Headers.ContentType = new MediaTypeHeaderValue("application/octet-stream");
43             var response = await client.PostAsync(Constants.PredictionUrl, content);
44
45             if (response.IsSuccessStatusCode)
46             {
47                 var predictionResult = await response.Content.ReadAsStringAsync();
48                 var customVisionResult = JsonConvert.DeserializeObject<CustomVisionResult>(predictionResult);
49
50                 return customVisionResult.Predictions.Where(x =>
51                     x.Score > Constants.Threshold).ToList();
52             }
53             else
54             {
55                 return new List<Prediction>();
56             }
57         }
58     }
59     catch (Exception ex)
60     {
61     }
62 }
```

At the bottom, the Breakpoints, Locals, Watch, and Threads toolbars are visible. To the right, an iPhone 11 simulator shows an image of a muffin with a bounding box drawn around it, with the word "Muffin" overlaid.

Proyecto open-source disponible en GitHub: <https://github.com/icebeam7/ObjectDetectionMobileApp>

# Tips de optimización

- La mejor forma de tener un detector de objetos de calidad es **agregando imágenes diversas** (diferentes fondos, ángulos, tamaño de objeto, grupos de fotos y variantes de tipos).
- **Incluye imágenes representativas de lo que el detector encontrará en el mundo real.** Por tanto, las fotos en contexto son mejores que las de objetos frente a fondos neutros, por ejemplo.
- Siempre entrena el modelo después de haber agregado más imágenes.
- Usa **al menos 30 imágenes** por etiqueta.



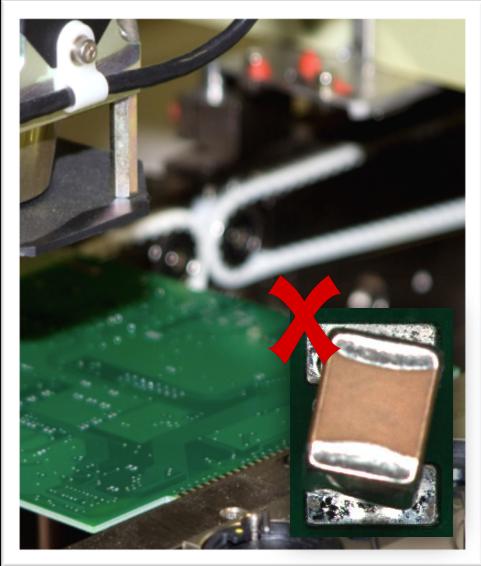
# Precio

## Pricing details

The below pricing reflects a preview discount.

INSTANCE	TRANSACTIONS PER SECOND (TPS)	FEATURES	PRICE
Free	2 TPS	Upload, training, and prediction transactions Up to 2 projects Up to 1 hour training per month	5,000 training images free per project 10,000 predictions per month
Standard	10 TPS	Upload and prediction transactions Up to 100 projects	\$2 per 1,000 transactions
		Training	\$20 per compute hour
		Image Storage Up to 6 MB each	\$0.70 per 1,000 images

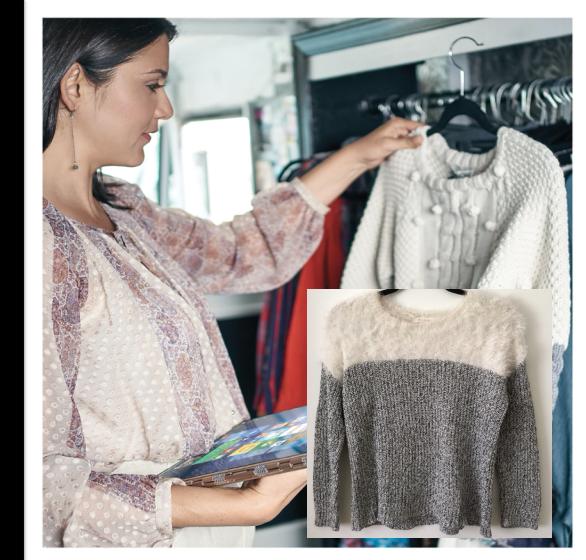
# Escenarios de uso



Detección de categoría en una línea de producción



Detección de la salud de plantas en agricultura y ganadería



Detección de productos en una tienda en línea

## Escenarios adicionales

- Clasificar imágenes enviadas por el usuario al sitio web
- Identificar elementos: conteo de objetos, identificación de animales, etc.
- Detección de riesgos / seguridad industrial

# Call to Action

## **Servicio de Custom Vision**

<https://azure.microsoft.com/es-mx/services/cognitive-services/custom-vision-service/>

## **Documentación de Custom Vision**

<https://docs.microsoft.com/es-mx/azure/cognitive-services/custom-vision-service/home>

## **Quickstart: Create an object detection project with the Custom Vision SDK**

<https://docs.microsoft.com/en-us/azure/cognitive-services/custom-vision-service/quickstarts/object-detection>

## **Custom Vision y TensorFlow**

<https://docs.microsoft.com/es-mx/azure/cognitive-services/custom-vision-service/export-model-python>

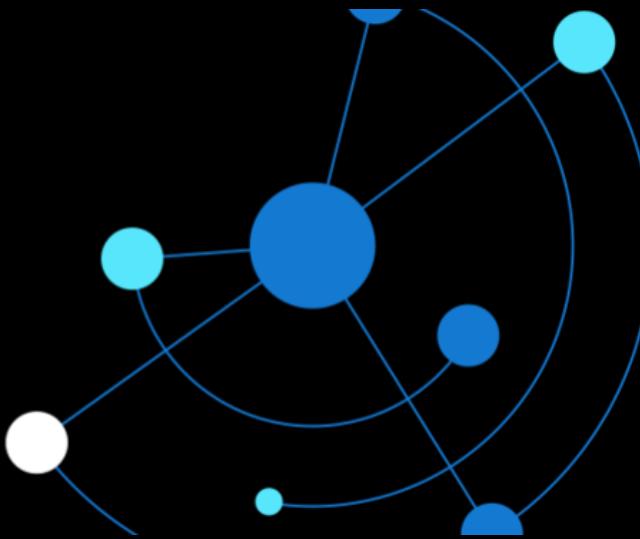


# Azure Tech Communities Live Days - Video Indexer & Azure Custom Vision

Combinando el poder de Video Indexer y Azure Custom Vision para la detección de objetos en videos con IA

Mayo 8 - 2020 / LATAM -

[Ver Agenda](#)



En este evento exploraremos juntos cómo combinar el poder de Video Indexer y Azure Custom Vision para la detección de objetos en videos implementando IA presentado por Luis Antonio Beltran – MVP Mexico (Desarrollador móvil certificado en Xamarin MS (C#, Azure), MCSD en UWP, MCPS, MTA).

Tendrás la oportunidad de preguntar todo lo que necesitas para continuar aprendiendo y elevando tu crecimiento profesional.

El evento comienza a las 02:00 PM EST (Eastern Standard Time). Los horarios locales por país son:

- 02:00 - 03:30 PM: República Dominicana, Puerto Rico, Trinidad y Tobago, Paraguay, Bolivia y Chile.
- 11:00 - 12:30 PM: Costa Rica, Guatemala, El Salvador, Honduras.
- 01:00 - 02:30 PM: Colombia, Jamaica, Ecuador, Perú, Panamá, México.
- 03:00 - 04:30 PM: Argentina, Uruguay.

## Agenda

**08/05**

Azure Tech Communities Live Days

02:00 PM Combinando el poder de Video Indexer y Azure Custom Vision para la detección de objetos en videos con IA

03:00 PM Preguntas y respuestas (Q&A)

<https://www.microsoft.com/es-xl/events-hub/latam/azure-tech-communities-live-days-video-indexer-amp-azure-custom-vision/>

## Global AI On Tour Latinoamérica (Online)

May 18th–23rd, 2020

Worldwide (Online event)

[View schedule](#)



Global AI On Tour Latinoamérica es un evento en línea dirigido a entusiastas de la Inteligencia Artificial en Microsoft Azure.

Aprende a implementar soluciones de IA utilizando servicios previamente entrenados, tales como los Cognitive Services y Bot Framework, o construye tus propios modelos de machine learning con Azure ML y frameworks de código abierto como PyTorch y ML.NET. Al final del día, ¡serás capaz de incorporar inteligencia en tus aplicaciones!

<https://ti.to/comunidad-xamarin-en-espanol/global-ai-on-tour-latinoamerica-2020>



Comunidad Xamarin en Español



Aprendiendo Azure

# ¡Gracias por tu atención!

**GitHub:**

<https://github.com/icebeam7>

**YouTube:**

<https://youtube.com/user/darkicebeam>

**About Me:**

<https://about.me/luis-beltran>

**LinkedIn:**

<https://linkedin.com/in/luisantoniobeltran>

**SlideShare:**

<https://slideshare.net/icebeam>

Luis Beltrán

Tomás Bata University in Zlín  
Tecnológico Nacional de México en Celaya



[luis@luisbeltran.mx](mailto:luis@luisbeltran.mx)

[luisbeltran.mx](http://luisbeltran.mx)

[@darkicebeam](https://@darkicebeam)

