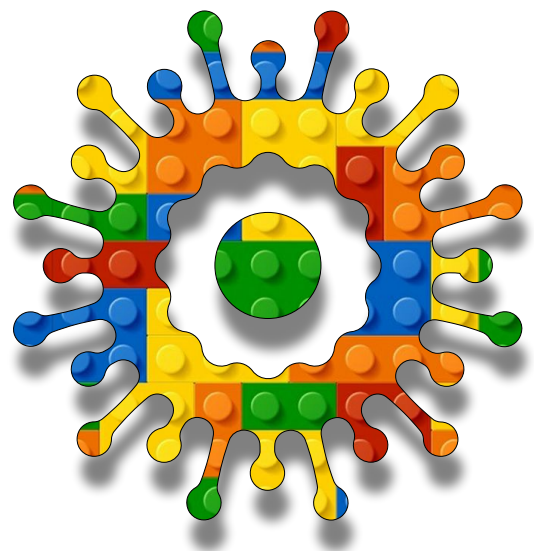


Algorithmic  
Intelligence  
Module B  
Unit #8  
mobileNet  
Video  
Classification





## Module B Unit #8 mobileNet video classification

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## Introduction to mobileNet video classification

Using **mobileNet** as before we are going to live stream and try to classify what it can see, should be fun! There are a lot of similarities to classifying an image.



## Sketch B8.1 index.html

Make sure you have the `ml5.js` line of code in your `index.html` file.

```
<!DOCTYPE html>
<html lang="en"><head>
  <script src="https://cdn.jsdelivr.net/npm/p5@2.2.2/lib/p5.js"></script>
  <script src="https://unpkg.com/ml5@1/dist/ml5.min.js"></script>
  <link rel="stylesheet" type="text/css" href="style.css">
  <meta charset="utf-8">
</head>
<body>
  <main>
  </main>
  <script src="sketch.js"></script>
</body></html>
```



## Sketch B8.2 starting sketch

Getting the video to work, you will get a prompt asking permission to allow access to the webcam.

```
let mobilenet
let video

async function setup()
{
  createCanvas(400, 400)
  video = await createCapture(VIDEO)
  video.hide()
  mobilenet = await ml5.imageClassifier('MobileNet')
}

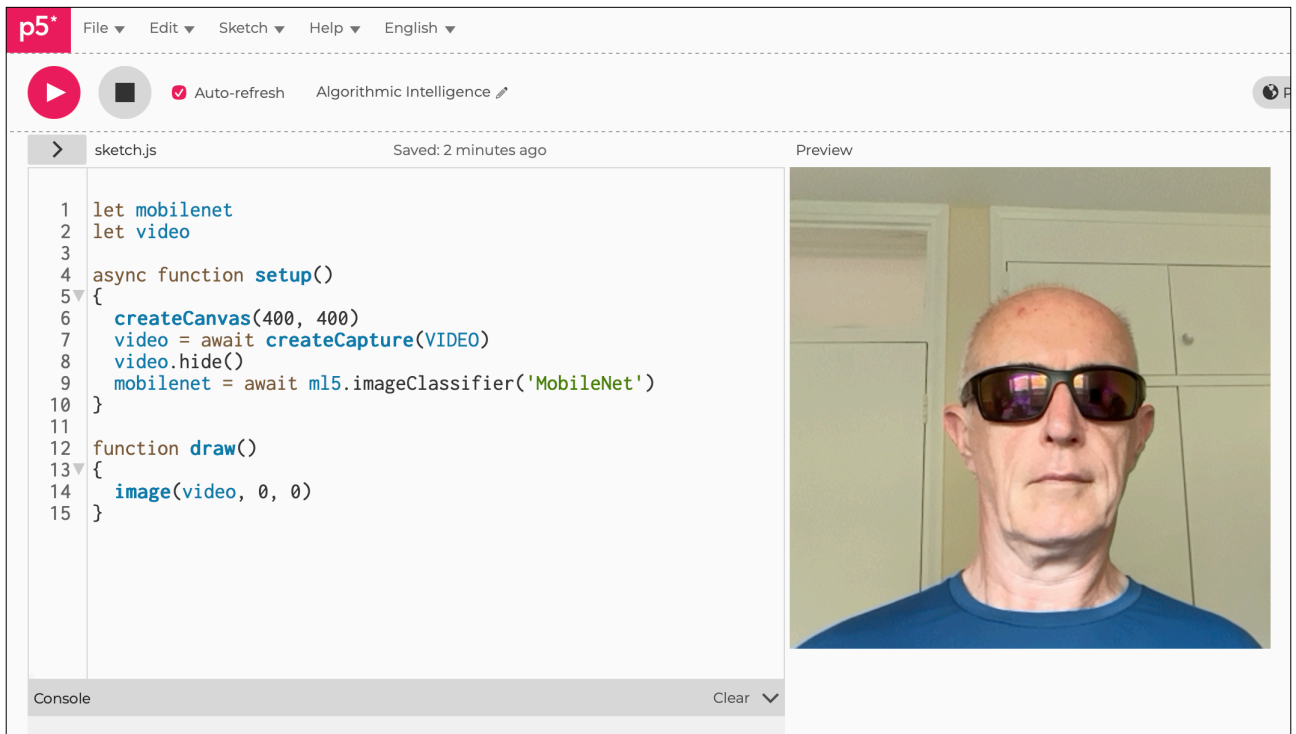
function draw()
{
  image(video, 0, 0)
}
```



### Notes

You should get an image of yourself like below, but maybe without the sunglasses. The reason I am wearing sunglasses will be explained shortly.

Figure B8.2





## Sketch B8.3 classifying

To classify a streaming video, we need to use a different function called `classifyStart()` rather than just `classify()`. We also add our old friend, the `gotResults()` function.

```
let mobilenet
let video

async function setup()
{
  createCanvas(400, 400)
  video = await createCapture(VIDEO)
  video.hide()
  mobilenet = await ml5.imageClassifier('MobileNet')
  mobilenet.classifyStart(video, gotResults)
}

function draw()
{
  image(video, 0, 0)
}

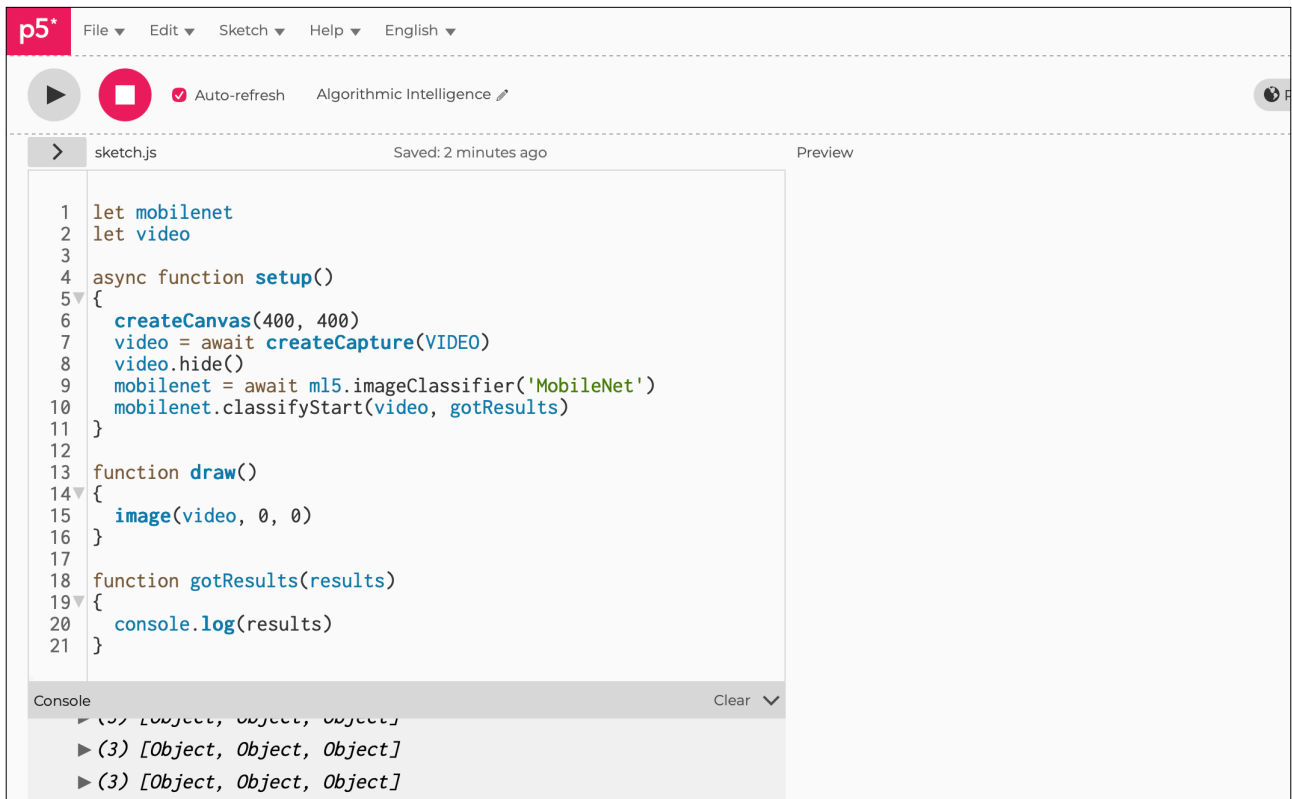
function gotResults(results)
{
  console.log(results)
}
```



### Notes

We get a continuous stream of array data, so press the stop/pause button and take a look inside one of these arrays. You will see three predictions with the usual label and confidence score. You might be very surprised by what it thinks it is looking at!

Figure B8.3





## Sketch B8.4 display the results

We can follow a similar trend as we did with the image, but to make sure we can see the results, I will put the text over a separate background.

```
let mobilenet
let video
let label = ''
let confidence = ''

async function setup()
{
  createCanvas(400, 400)
  video = await createCapture(VIDEO)
  video.hide()
  mobilenet = await ml5.imageClassifier('MobileNet')
  mobilenet.classifyStart(video, gotResults)
}

function draw()
{
  image(video, 0, 0)
  fill(255)
  rect(0, height - 50, width, 50)
  textSize(20)
  fill(0)
  text(label, 10, height - 30)
  text(floor(confidence * 100) + '%', 10, height - 8)
}

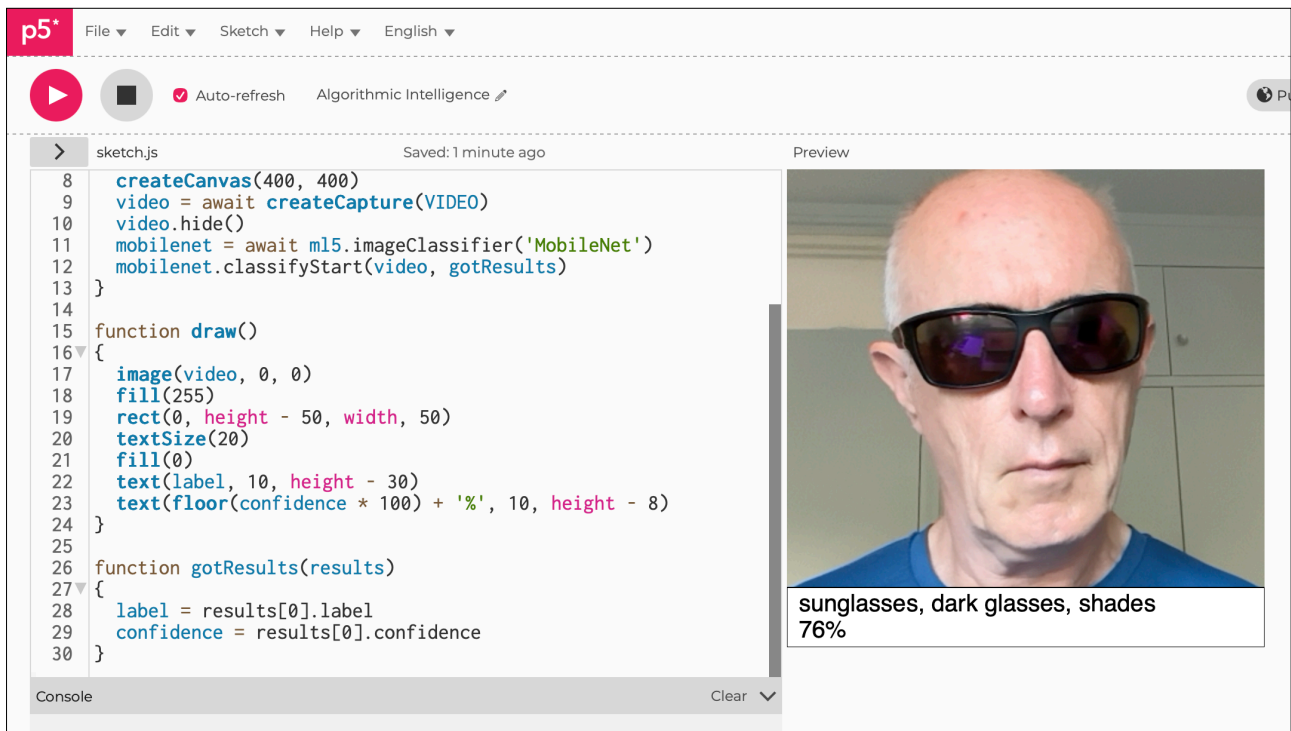
function gotResults(results)
{
  label = results[0].label
  confidence = results[0].confidence
}
```



### Notes

This is why I wore the sunglasses; it gave a very correct answer, even if the confidence value varied from 30% to 80% constantly. Some of the other predictions were a lot less flattering, to say the least.

Figure B8.4





## Sketch B8.5 a bit of a mug

By adding a button and changing the predicting function from `classifyStart()` to `classify()` and then putting it in another function called `snapshot()`, we can now position something and classify it just once.

```
let mobilenet
let video
let label = ''
let confidence = ''
let button

async function setup()
{
  createCanvas(400, 400)
  video = await createCapture(VIDEO)
  video.hide()
  mobilenet = await ml5.imageClassifier('MobileNet')
  // mobilenet.classifyStart(video, gotResults)
  button = createButton('take a snapshot')
  button.mousePressed(snapshot)
}

function snapshot()
{
  mobilenet.classify(video, gotResults)
}

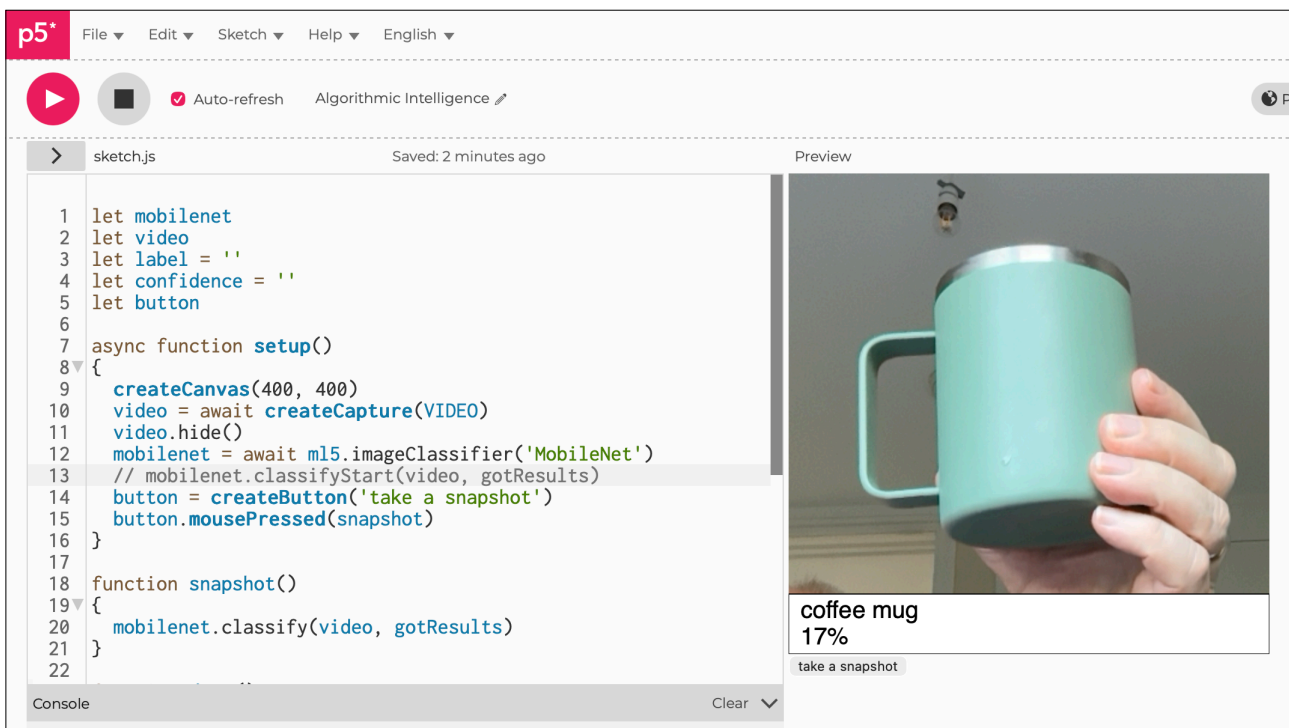
function draw()
{
  image(video, 0, 0)
  fill(255)
  rect(0, height - 50, width, 50)
  textSize(20)
  fill(0)
  text(label, 10, height - 30)
  text(floor(confidence * 100) + '%', 10, height - 8)
}
```

```
function gotResults(results)
{
  label = results[0].label
  confidence = results[0].confidence
}
```

## Notes

This is repeatable, so get another object and click on the button.

Figure B8.5



The screenshot shows the p5.js IDE interface. The code editor on the left contains the following JavaScript code:

```
1 let mobilenet
2 let video
3 let label = ''
4 let confidence = ''
5 let button
6
7 async function setup()
8 {
9   createCanvas(400, 400)
10  video = await createCapture(VIDEO)
11  video.hide()
12  mobilenet = await ml5.imageClassifier('MobileNet')
13  // mobilenet.classifyStart(video, gotResults)
14  button = createButton('take a snapshot')
15  button.mousePressed(snapshot)
16 }
17
18 function snapshot()
19 {
20   mobilenet.classify(video, gotResults)
21 }
22
```

The preview window on the right shows a hand holding a teal coffee mug. Below the image, the text "coffee mug" is displayed with a confidence of "17%". A button labeled "take a snapshot" is visible at the bottom of the preview area.