

# Bottle, Can or Coffee Cup ?!



How Computer Vision and Machine Learning can be used to Recognise Different Materials to Make Recycling Easier



## 04 Improving the Performance for New Objects

This Materials Made Smarter Outreach Demonstration of How Computer Vision and Machine Learning can be used to Recognise Different Materials to Make Recycling Easier has been developed by Dr Robert Gibbs with Professor Cinzia Giannetti of Swansea University [\[↵\]](#) for Materials Made Smarter [\[↵\]](#), based upon the NVIDIA DLI "Getting Started with AI on Jetson Nano" course [\[↵\]](#).

This guide describes re-training the model to improve performance on examples of cans, bottles and coffee cups it has not seen before. An accompanying walkthrough video is available at Discover Materials by scanning the QR code or at <https://discovermaterials.co.uk/resource/bottle-can-or-coffee-cup/>

The video forms part of the section

### 04 Training for Different Materials

A playlist of all 4 videos is at

[https://www.youtube.com/playlist?list=PLyl3ubsSP6pUkBdTephBtqL7UfIFfGQ\\_Z](https://www.youtube.com/playlist?list=PLyl3ubsSP6pUkBdTephBtqL7UfIFfGQ_Z)

Also available on the Discover Materials website are a **glossary** of the **highlighted technical terms**, an electronic version of the printed **booklet** and further information about the code, the equipment and progressively more detailed project documentation.



Y Gyfadran Gwyddoniaeth a Pheirianneg  
Faculty of Science and Engineering

Materials and Manufacturing Research Institute



Engineering and Physical Sciences Research Council

developed by Dr R. Gibbs and Prof. C. Giannetti for Materials Made Smarter, based upon the NVIDIA DLI "Getting Started with AI on Jetson Nano" course. C.G. would like to acknowledge the support of the EPSRC (EP/V061798/1).



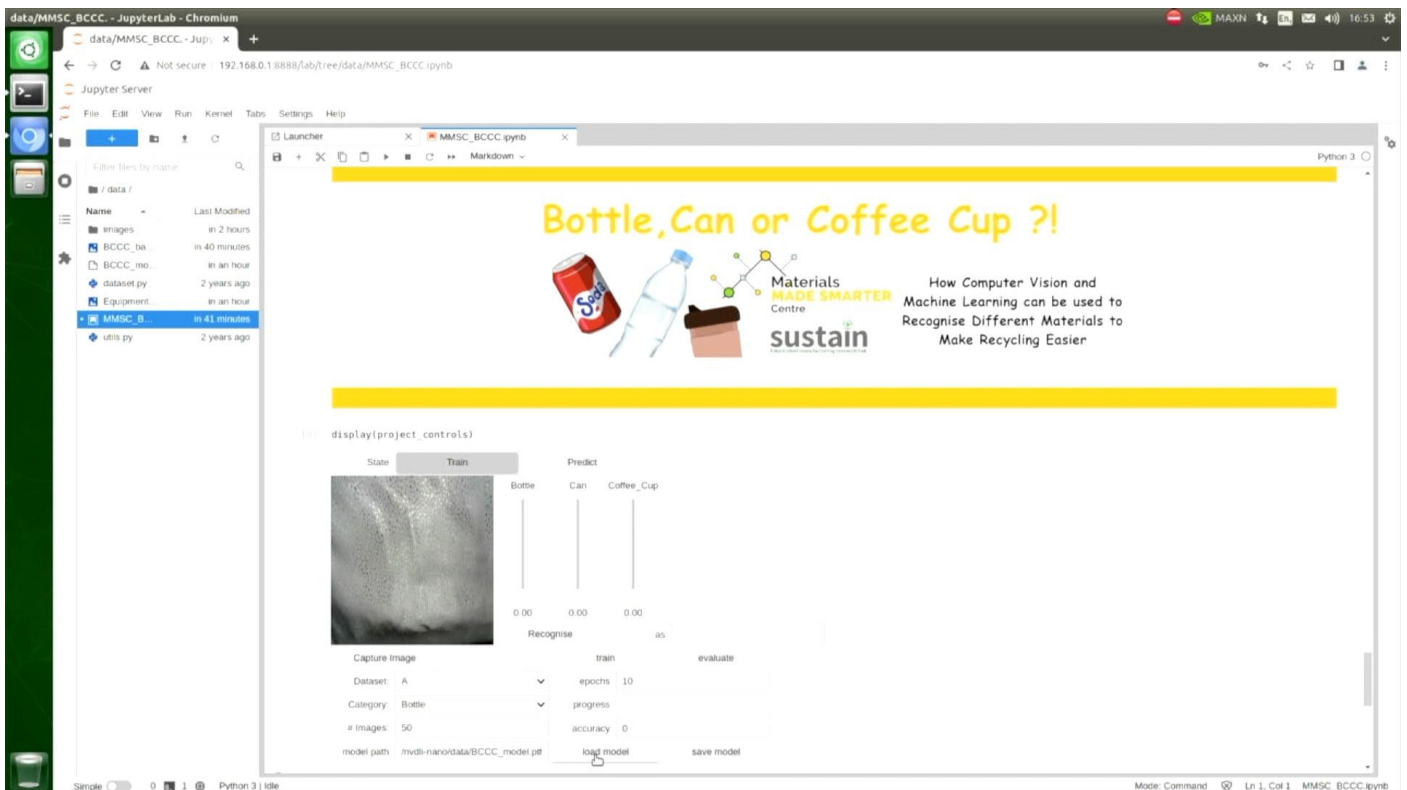
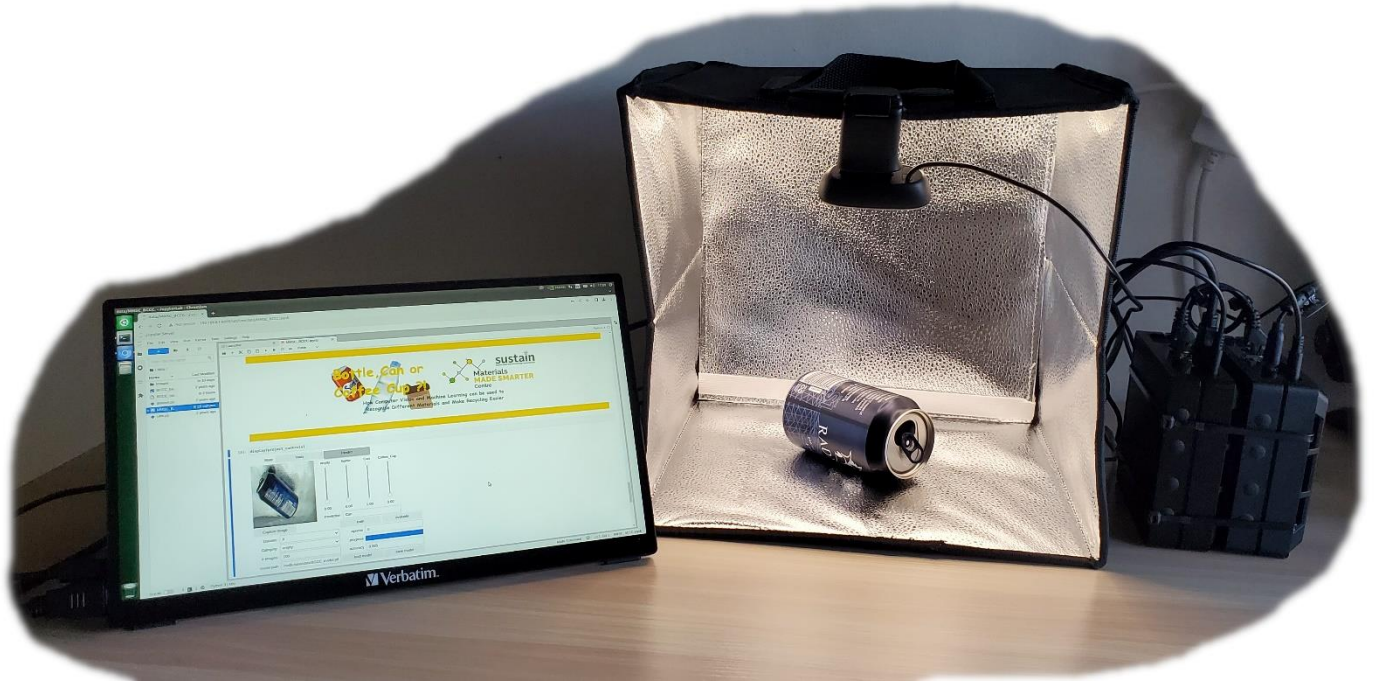
DEEP LEARNING INSTITUTE

visit **DISCOVER MATERIALS**

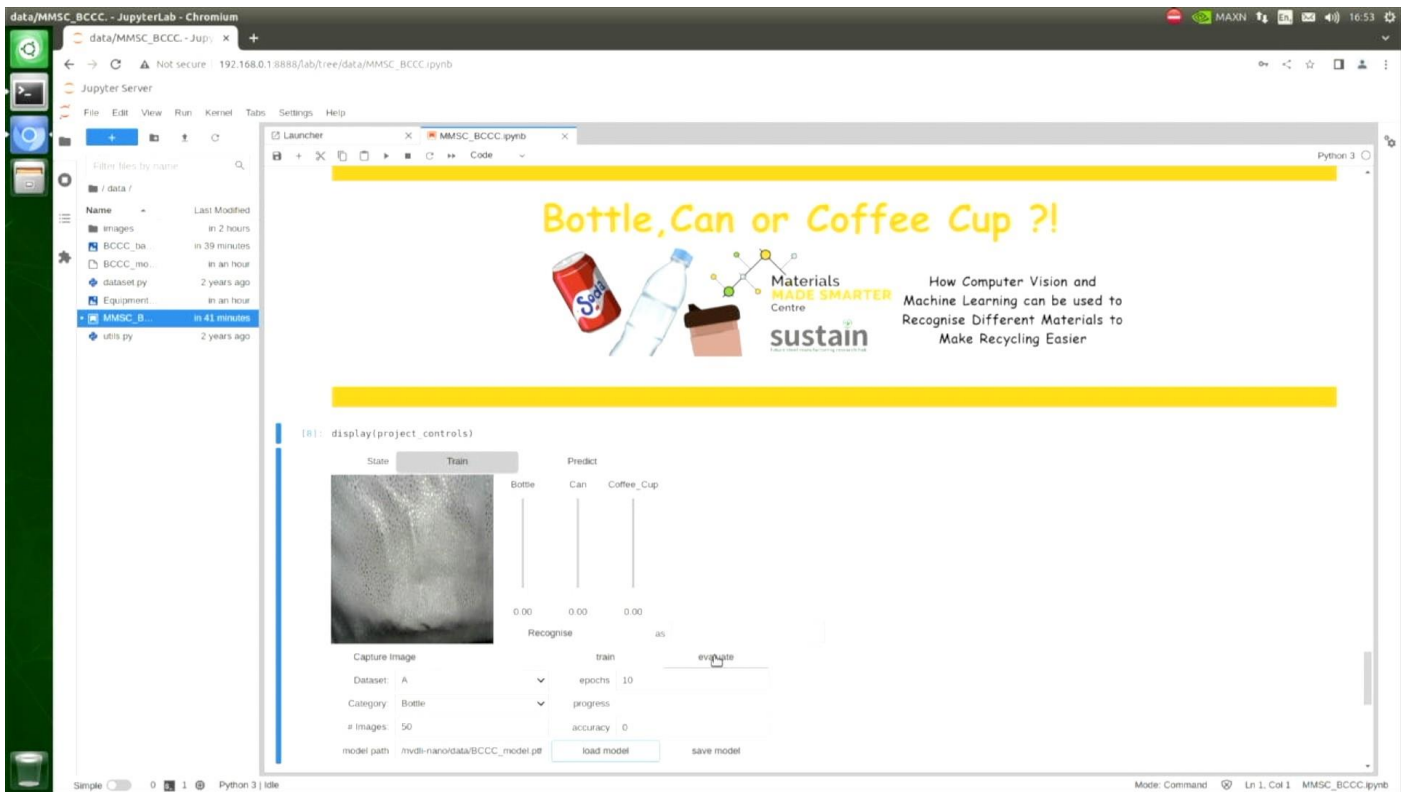
<https://discovermaterials.co.uk>

and learn more about what's happening in the world of materials science!

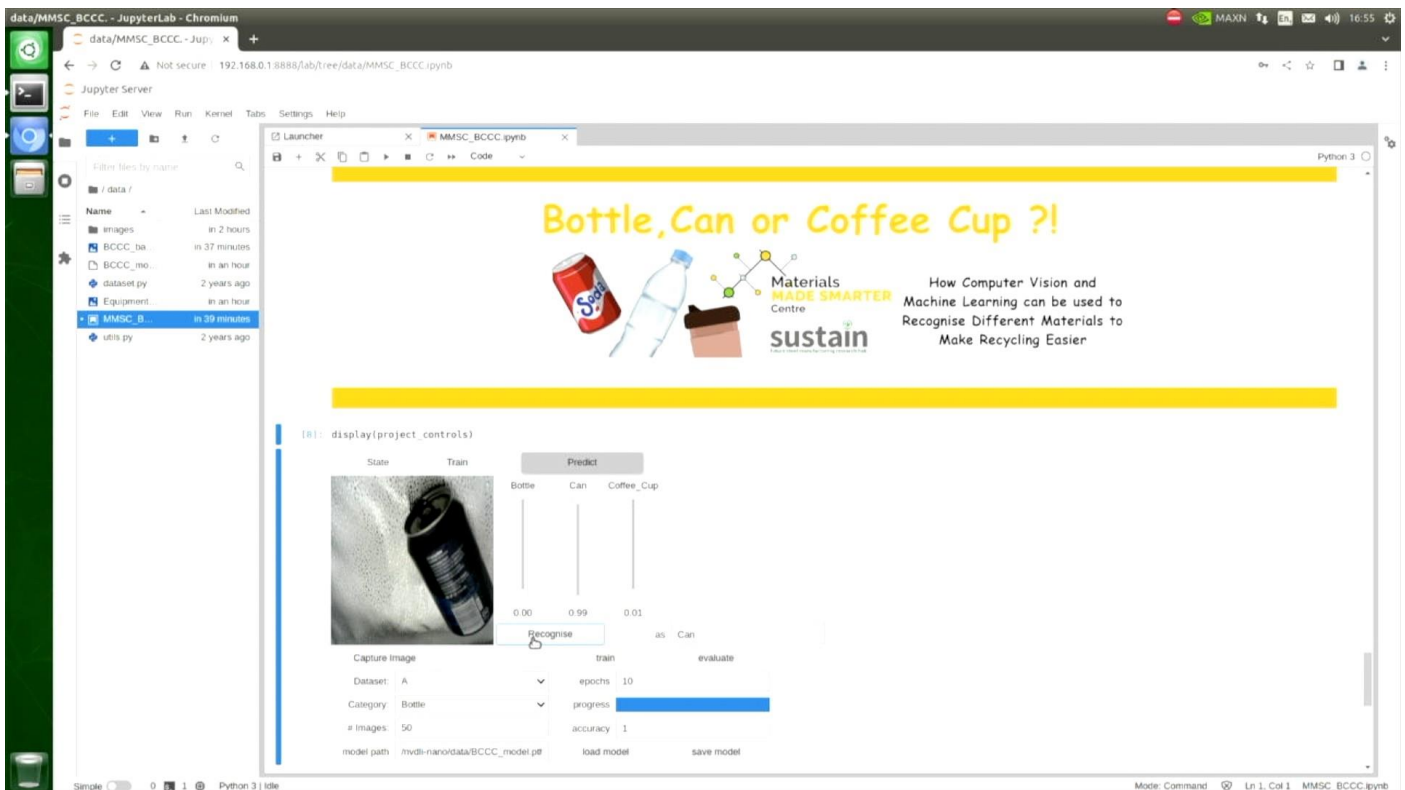
It is assumed that you have the system up and running according to the guide [01 Getting the System Up and Running](#) available on the website.



We've previously saved our trained model (see the guide [02 How to Use the Project Controls to Identify Objects](#), available on the website, for the process of training the model) so we don't need to re-run our training. We can press **load model** to load in our saved model.

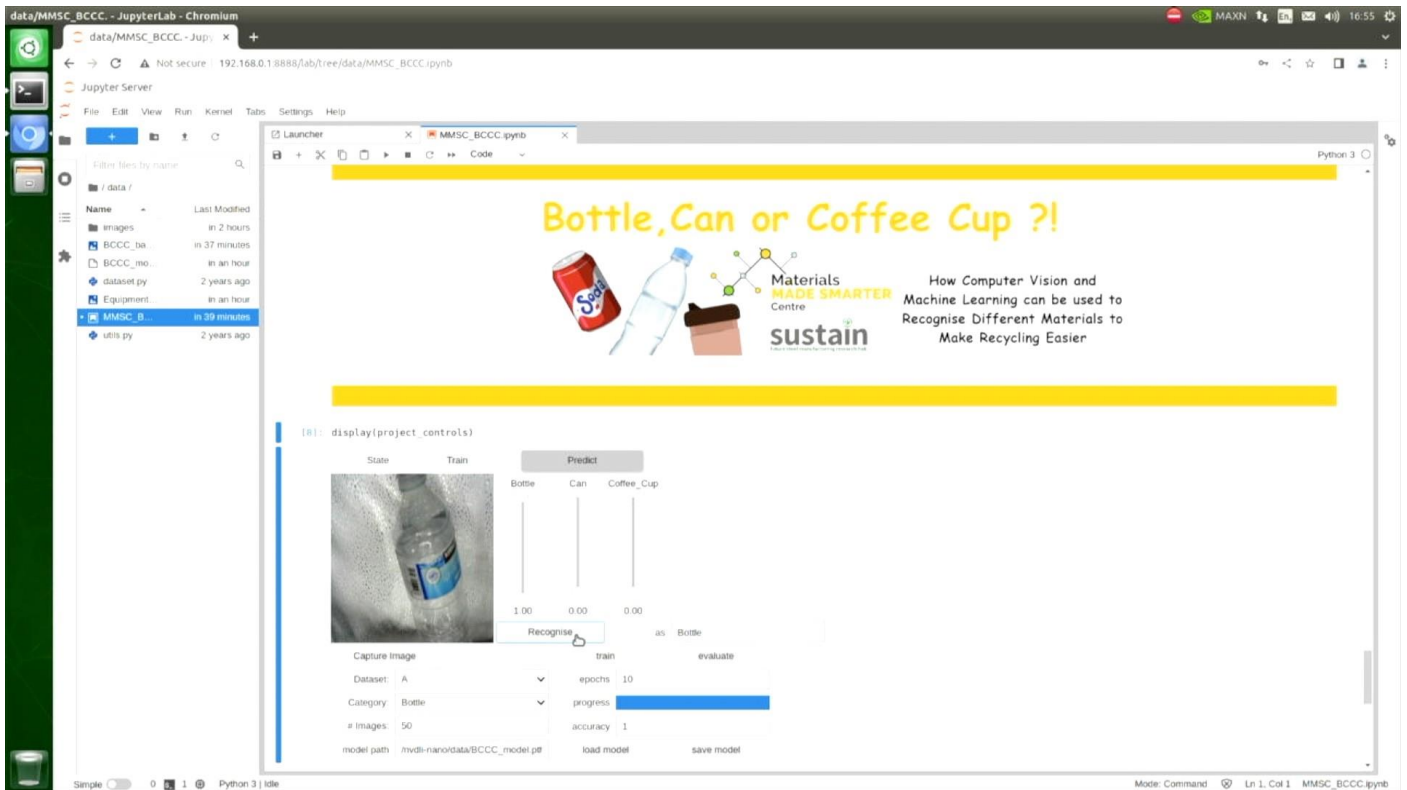


Instead of training we **evaluate** the model that we've just loaded in, This still takes approximately 60 seconds because it still has to load in the ResNet 18 model and replace the final layer of that model but it is much faster than the 4.6 minutes it takes to train from scratch with our existing training dataset.

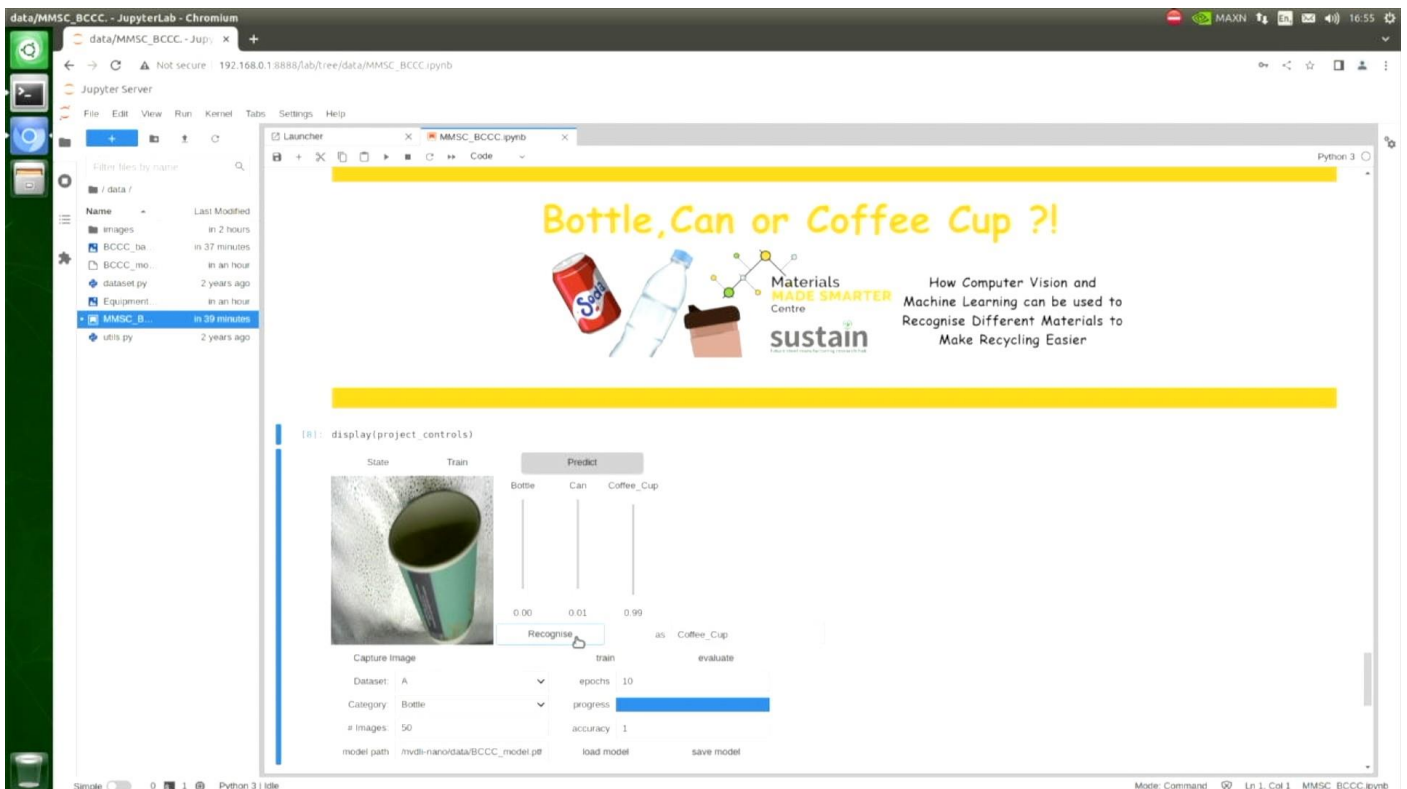


The system again switches to the predict mode, ready to recognize the objects you put in

It recognises the can with 99% confidence.

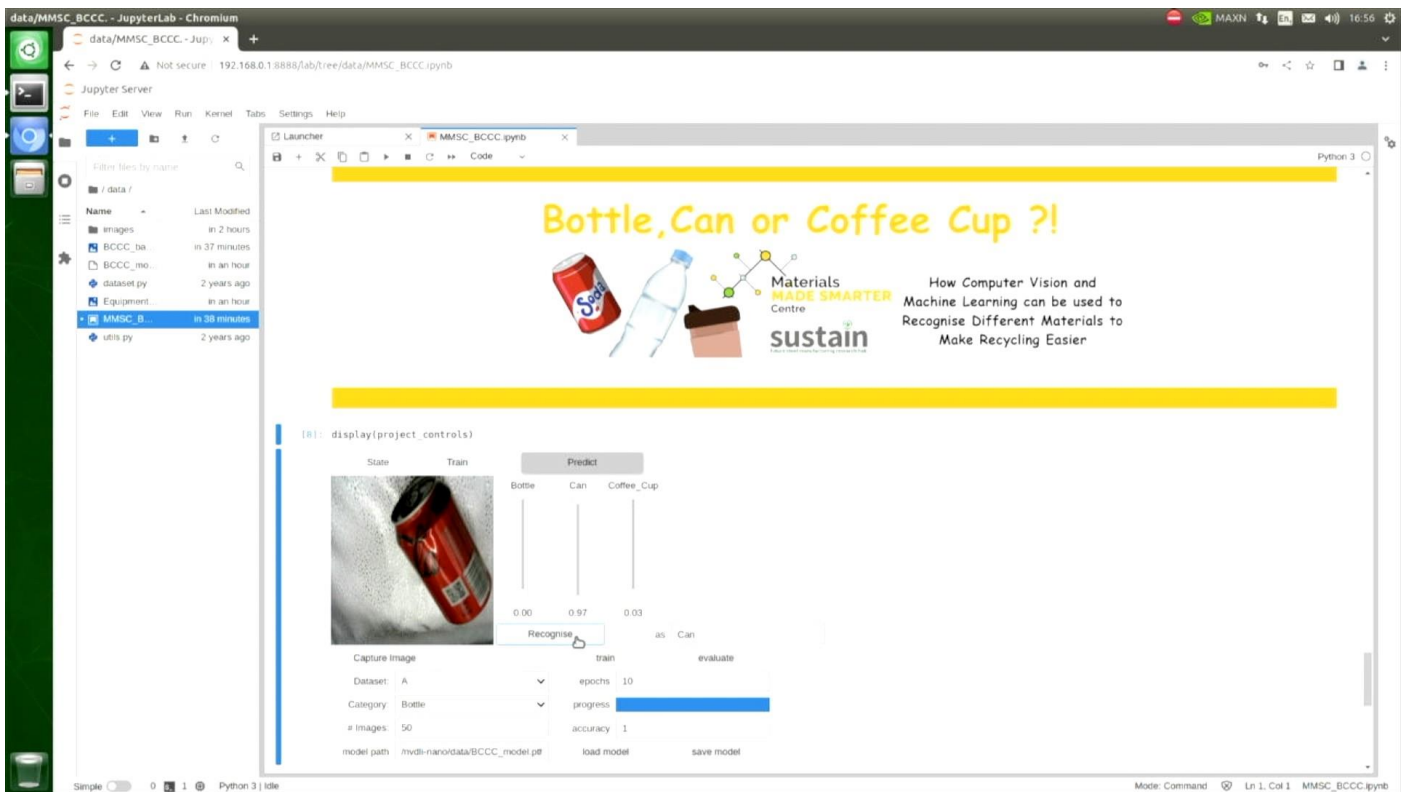


It recognizes the bottle with 100% confidence

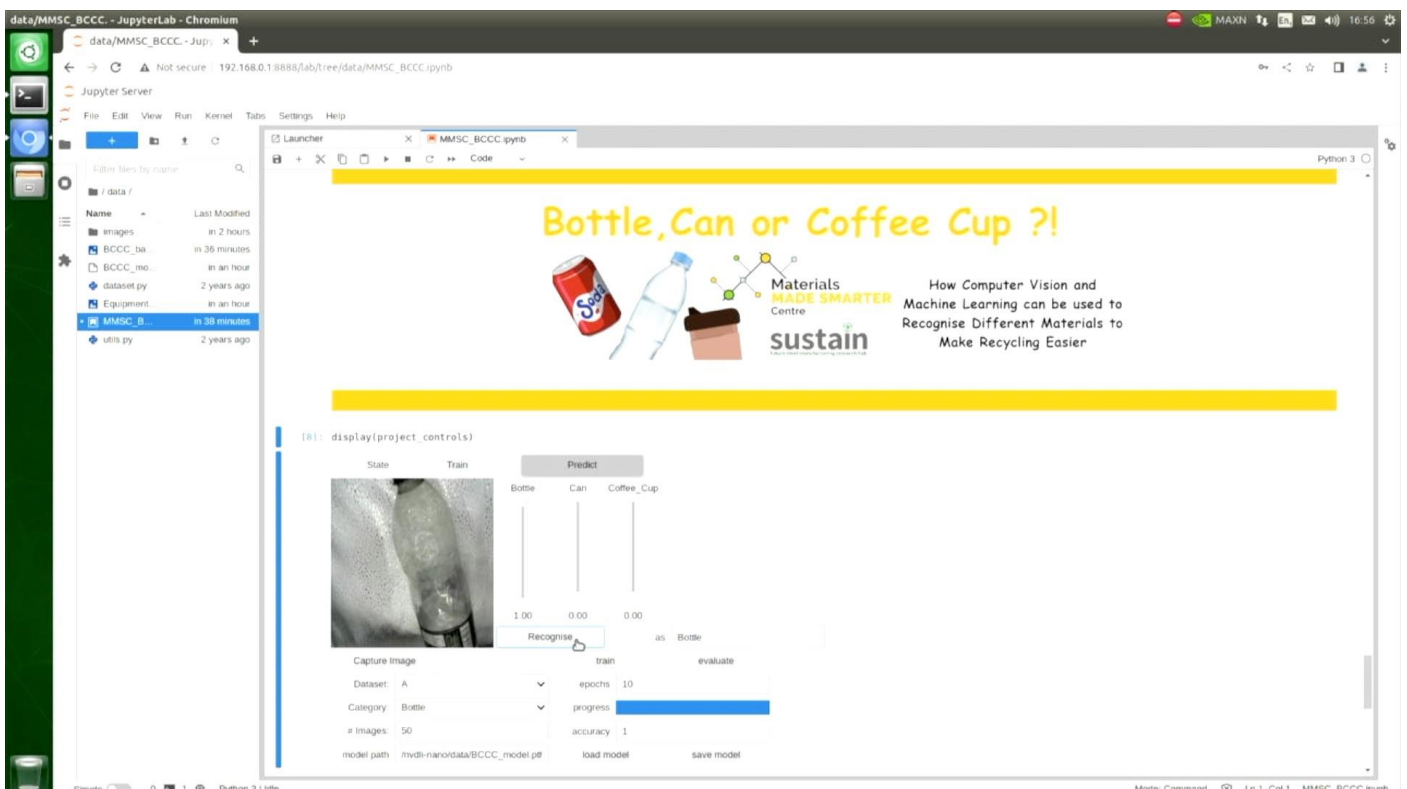


and also the coffee cup it has been trained on with 99% confidence.

Thumbnails of the original training dataset of 50 images of these three objects, stored in **BCCC\_A** are attached to the end of this guide.

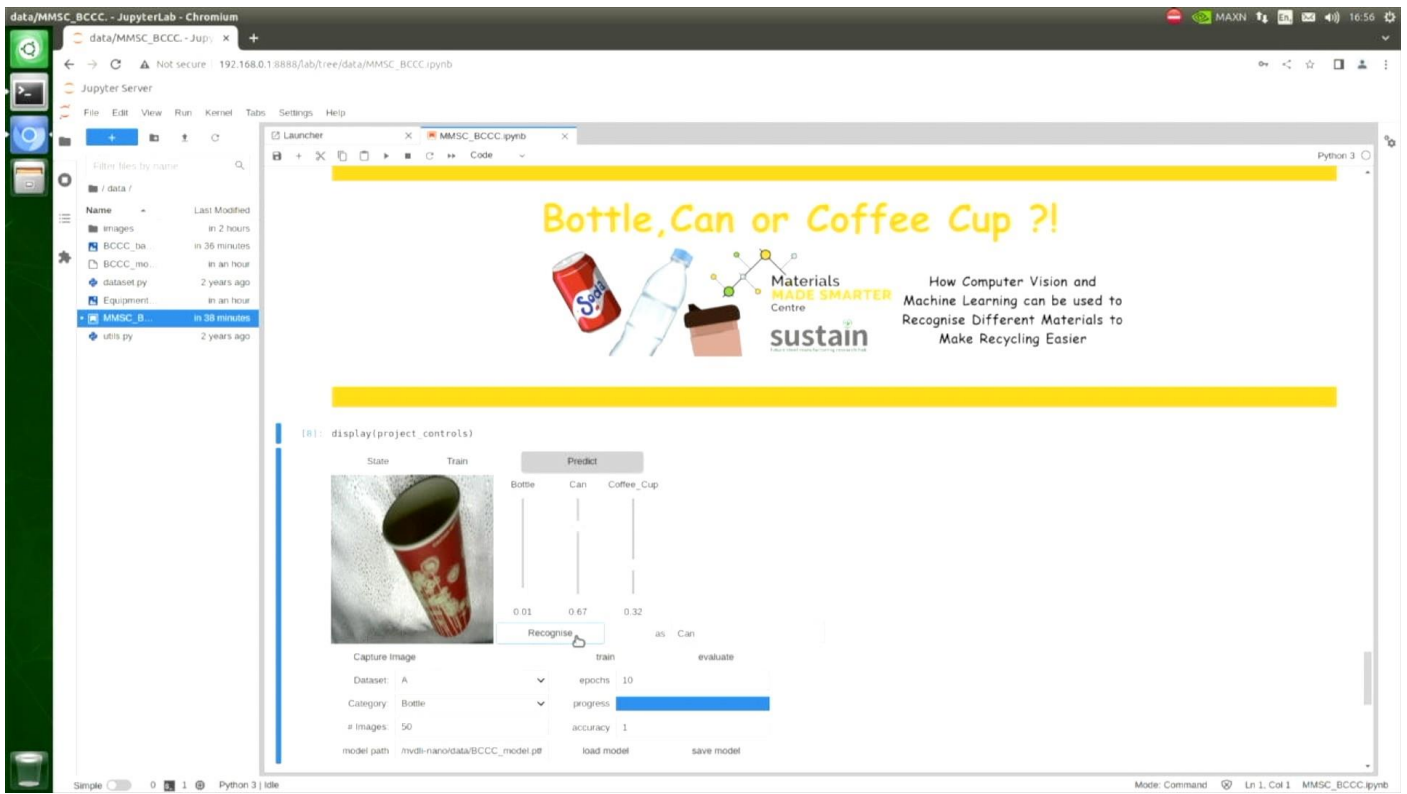


If we experiment with putting a can that the system hasn't seen before it successfully recognizes another can. There something common to both cans, even though they have different colors. The system has only 97% confidence, but that still very confident.



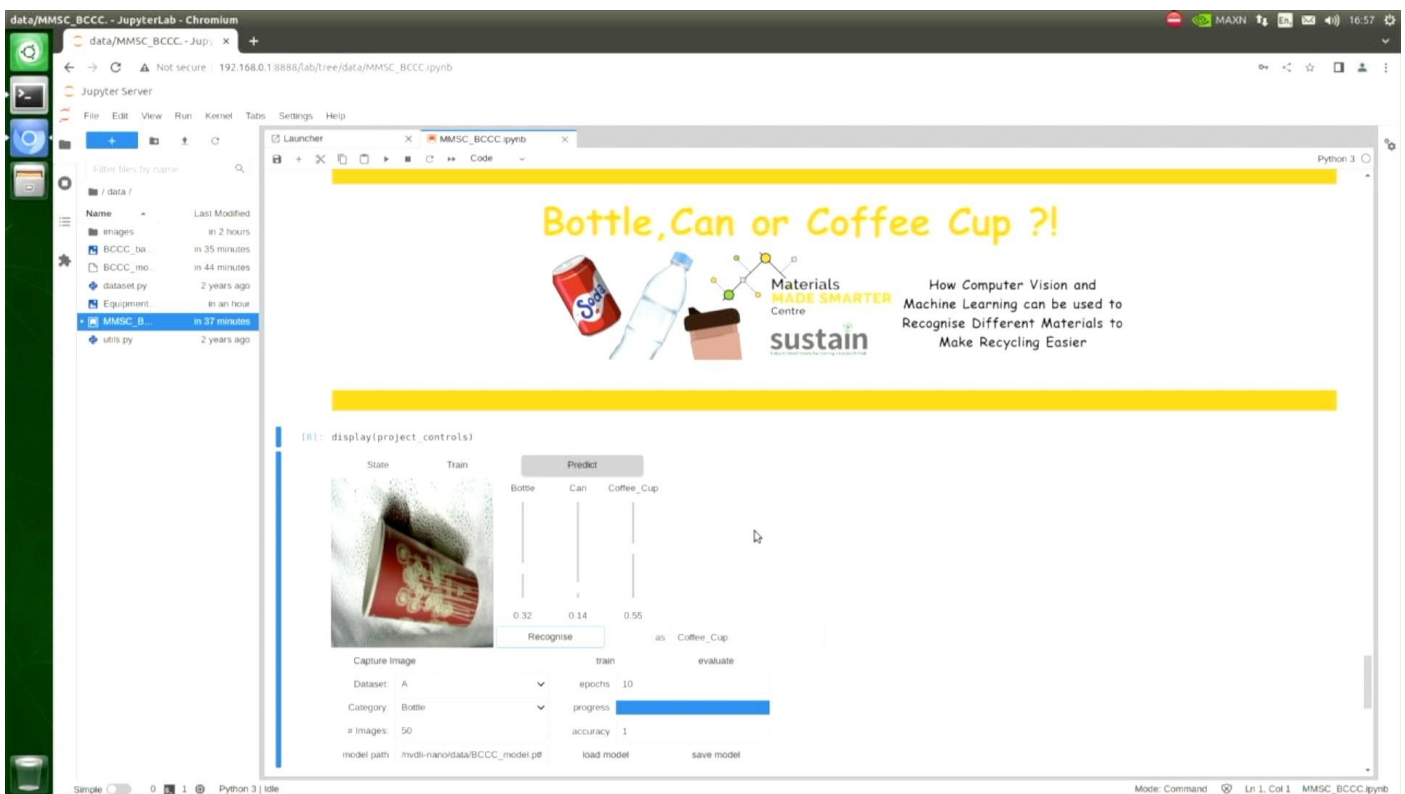
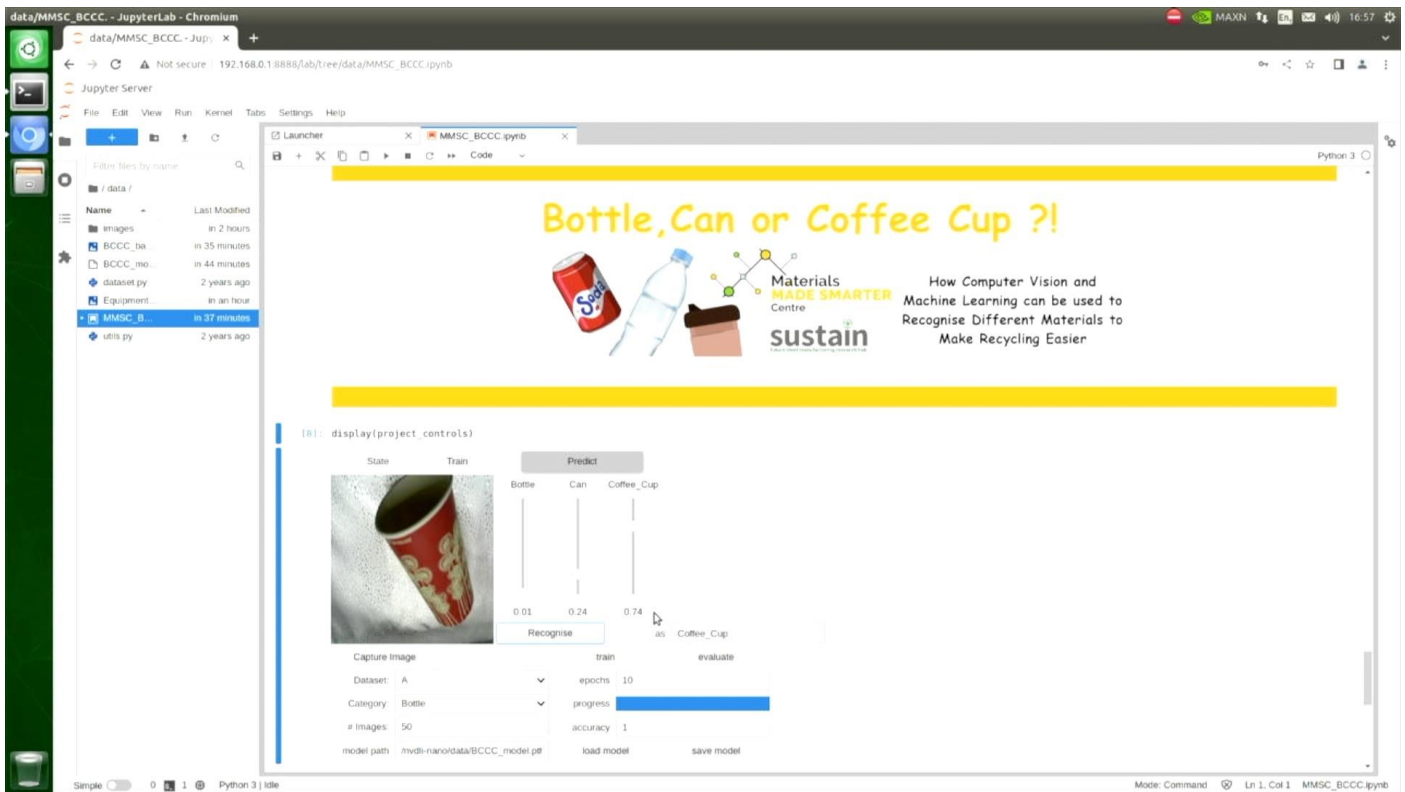
A bottle it hasn't been trained on can be identified with 100% confidence.

The system has a good understanding of the features of a bottle. What do you think is common in bottles with is different to the other three objects that they are easy to recognize?



Now if we try a coffee cup that is different to the one it has been trained on the system is confused about what the object is

There must be something in the coffee cup that it has been trained on, that it recognizes happily, that isn't in the new coffee cup.



Sometimes the system does identify the coffee cup. But only with 74% or 55% confidence

This coffee cup is a new coffee cup it hasn't seen before and it can easily be confused.

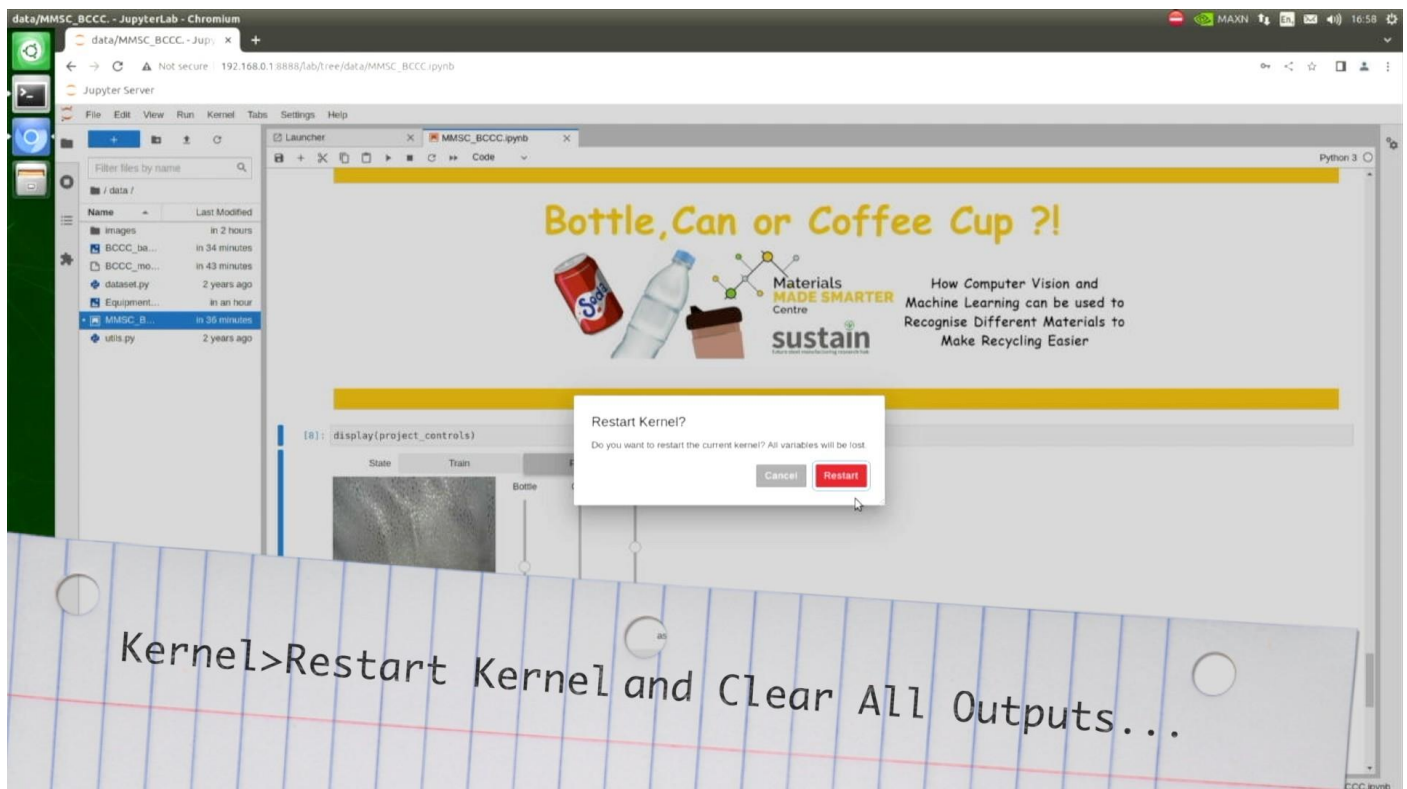
So how can we improve the situation on objects it hasn't seen before?

The key is giving it some more training images on the new versions of the objects so it can get a more general impression of what a coffee cup is, and be less confused.

What we want to do is just give it a few extra images of what a coffee cup looks like, and maybe also what the new can and the bottle looks like, but we don't want to damage our original data sets?

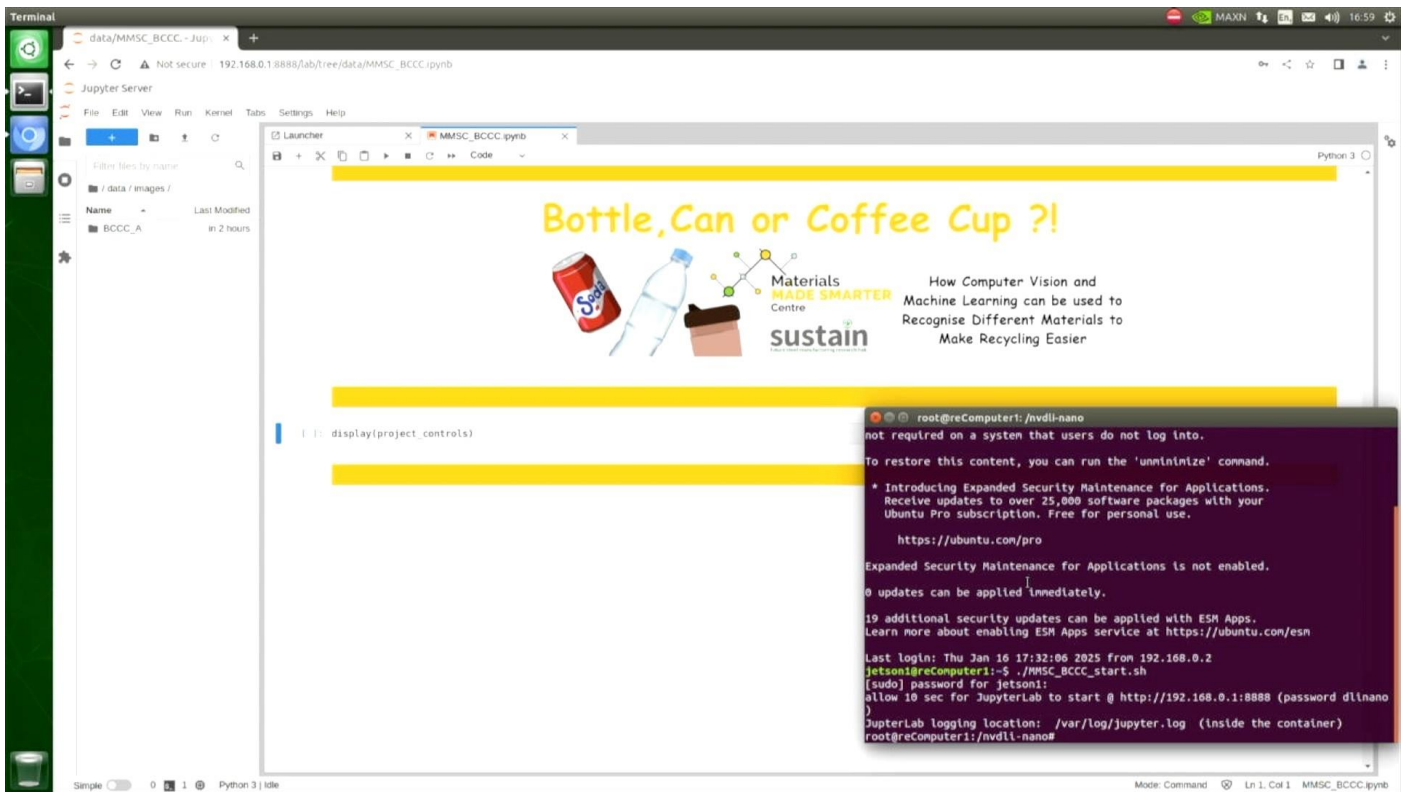
We want to keep **BCCC\_A** intact but we don't want to have to repeat the capture of the 50 images we have already obtained.

We are going to use the terminal to make an alternative copy of the original dataset, **BCCC\_B**, which we can add more images to.



We need to stop the system from running so we go to **Kernel>Restart Kernel and Clear All Outputs...** and click on **Restart** when it appears.

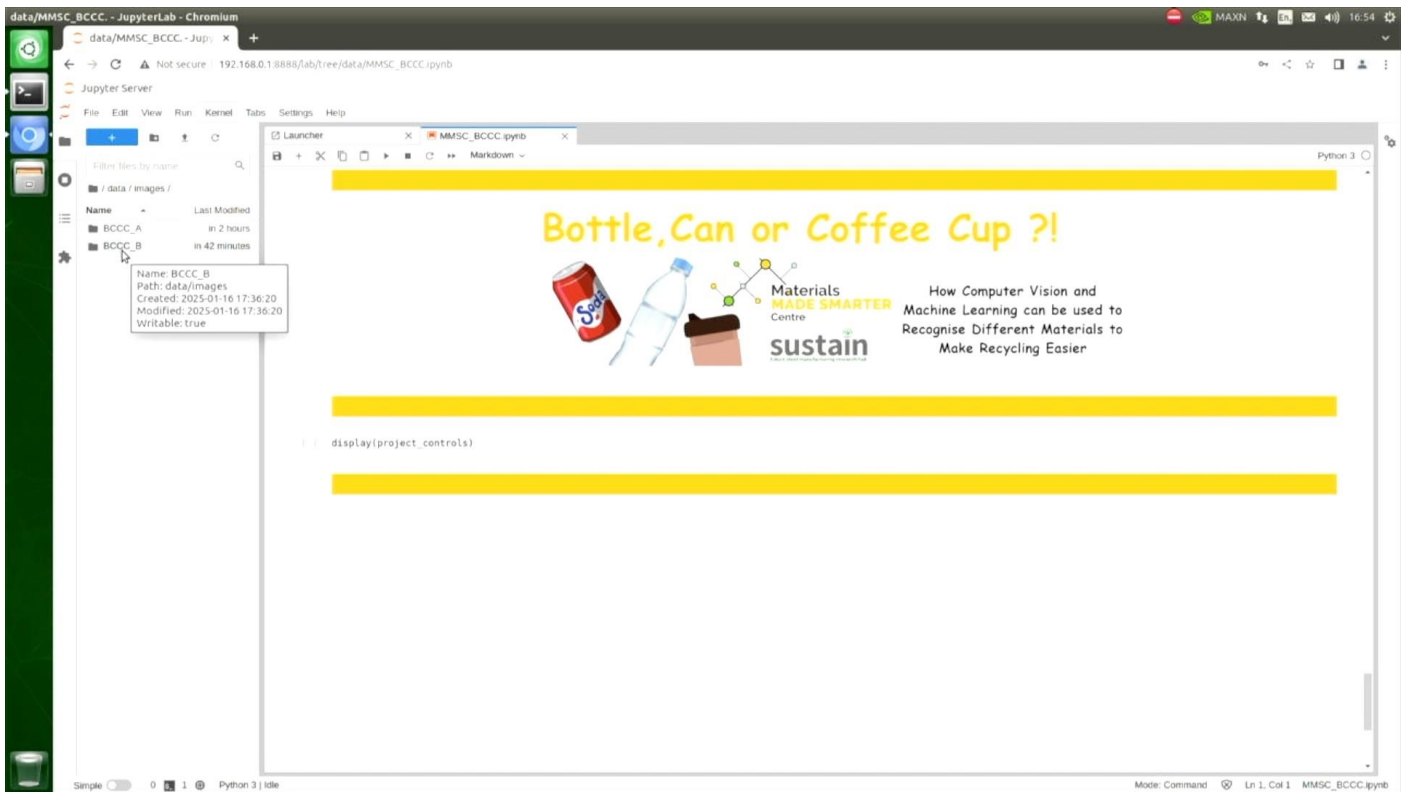




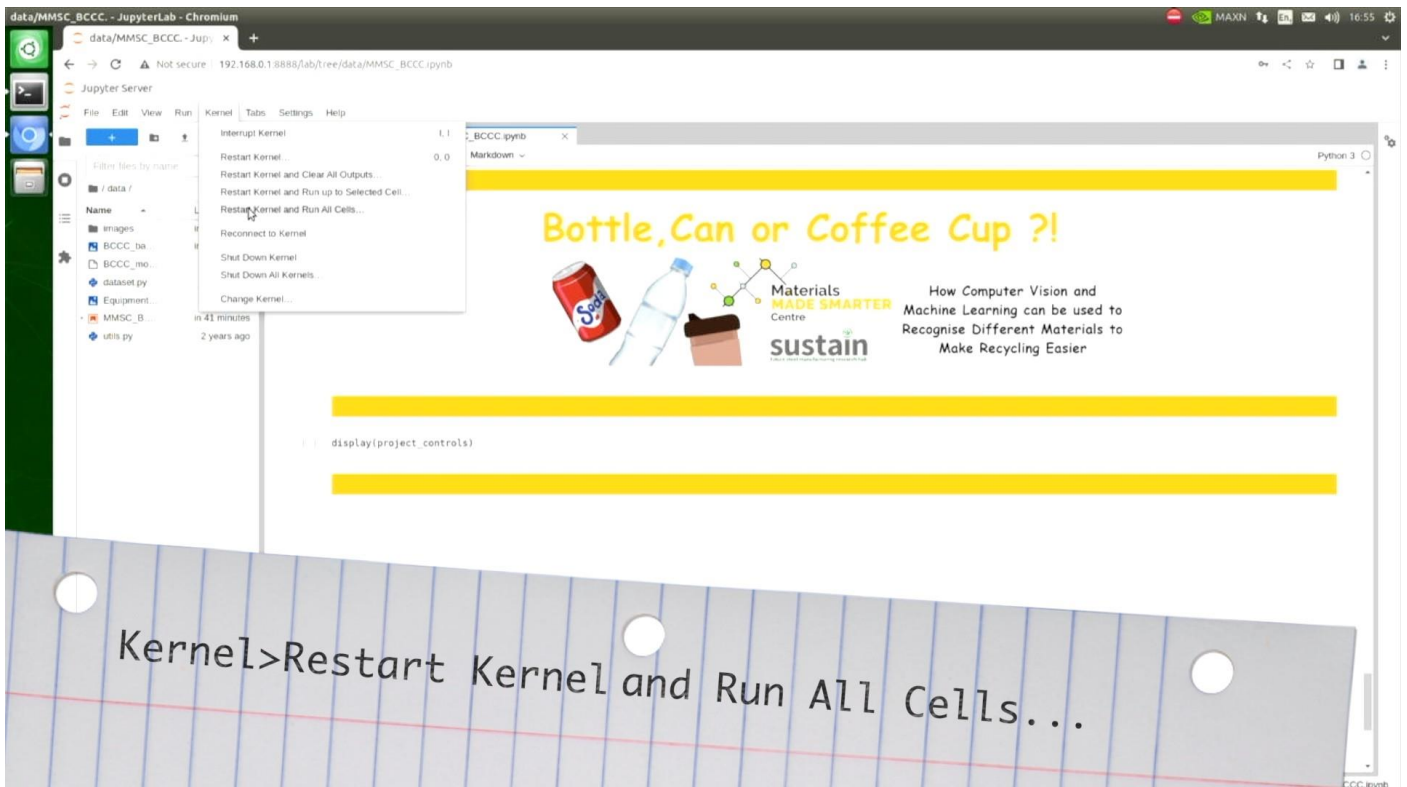
We click onto the terminal

In the terminal we will use the following commands in order to copy the existing BCCC\_A folder into a new BCCC\_B folder, pressing return after each step.

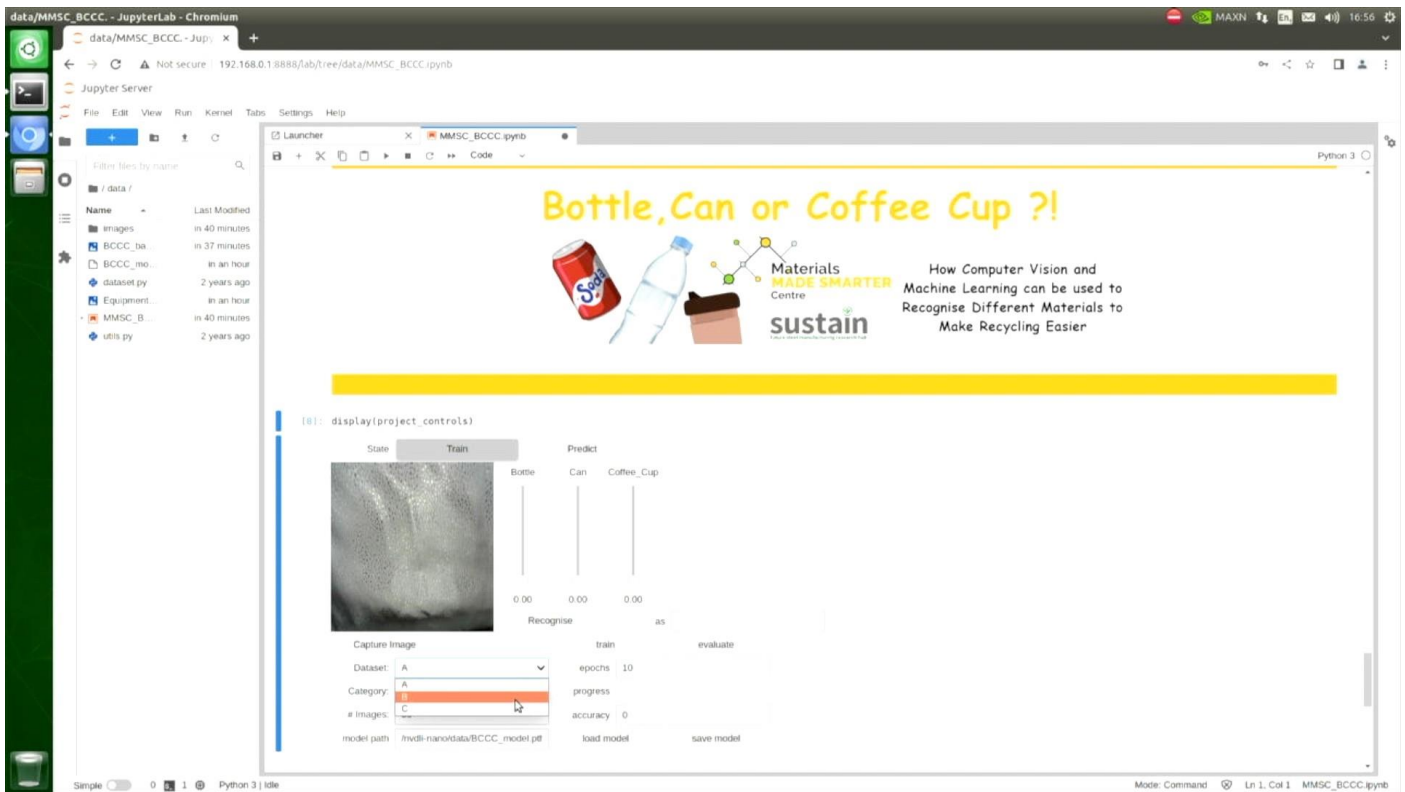
```
root@reComputer1:/nvdli-nano# ls
root@reComputer1:/nvdli-nano# cd data
root@reComputer1:/nvdli-nano# ls
root@reComputer1:/nvdli-nano# cd images
root@reComputer1:/nvdli-nano# ls
root@reComputer1:/nvdli-nano# cp -r ./BCCC_A/ ./BCCC_B/
root@reComputer1:/nvdli-nano# ls
```



and then we can return to the Jupyter Notebook in the chromium browser and see that our replicated BCCC\_B dataset is present, already populated with 50 images of the original objects.

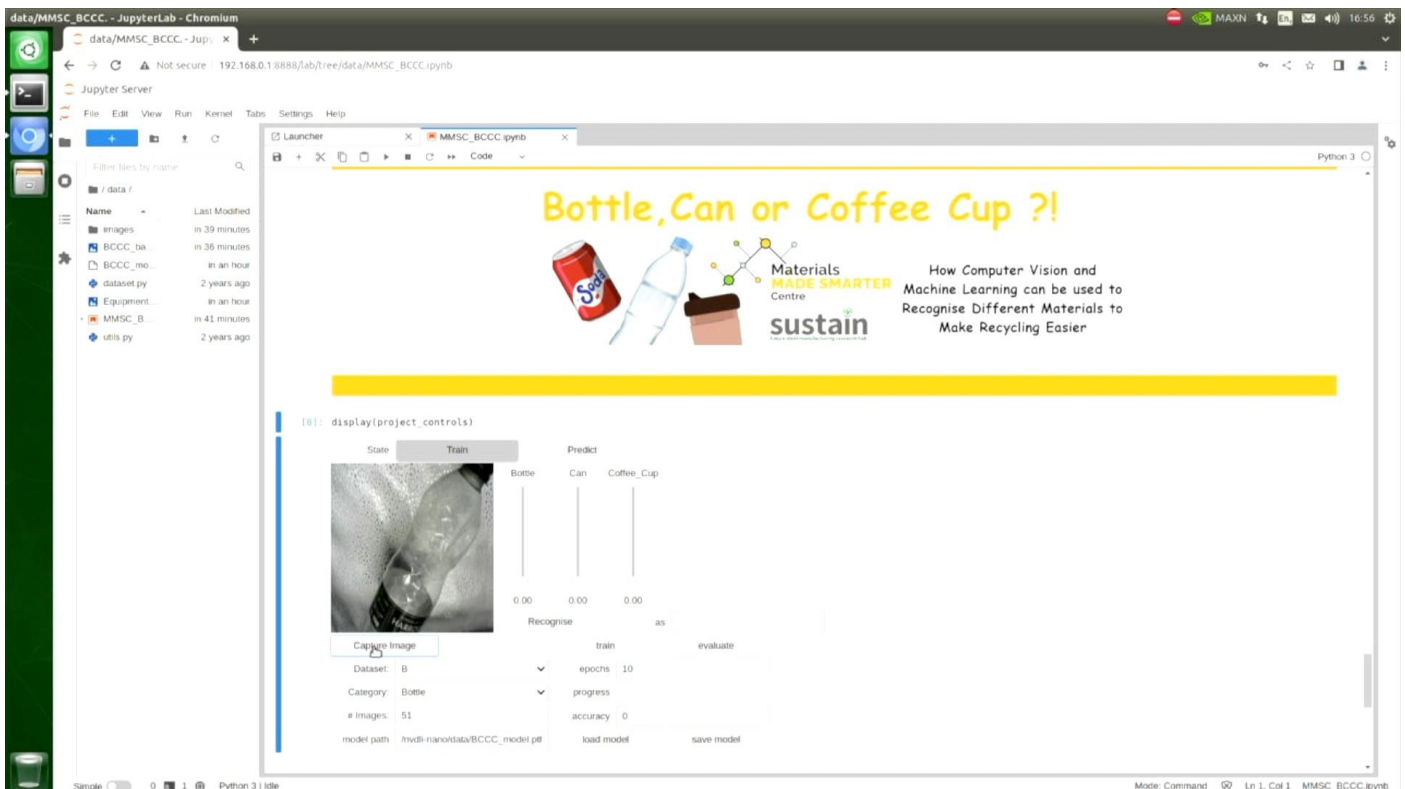


We go to **Kernel>Restart Kernel and Run All Cells...** and click on **Restart** when it appears to restart the system. Wait for 50 seconds.

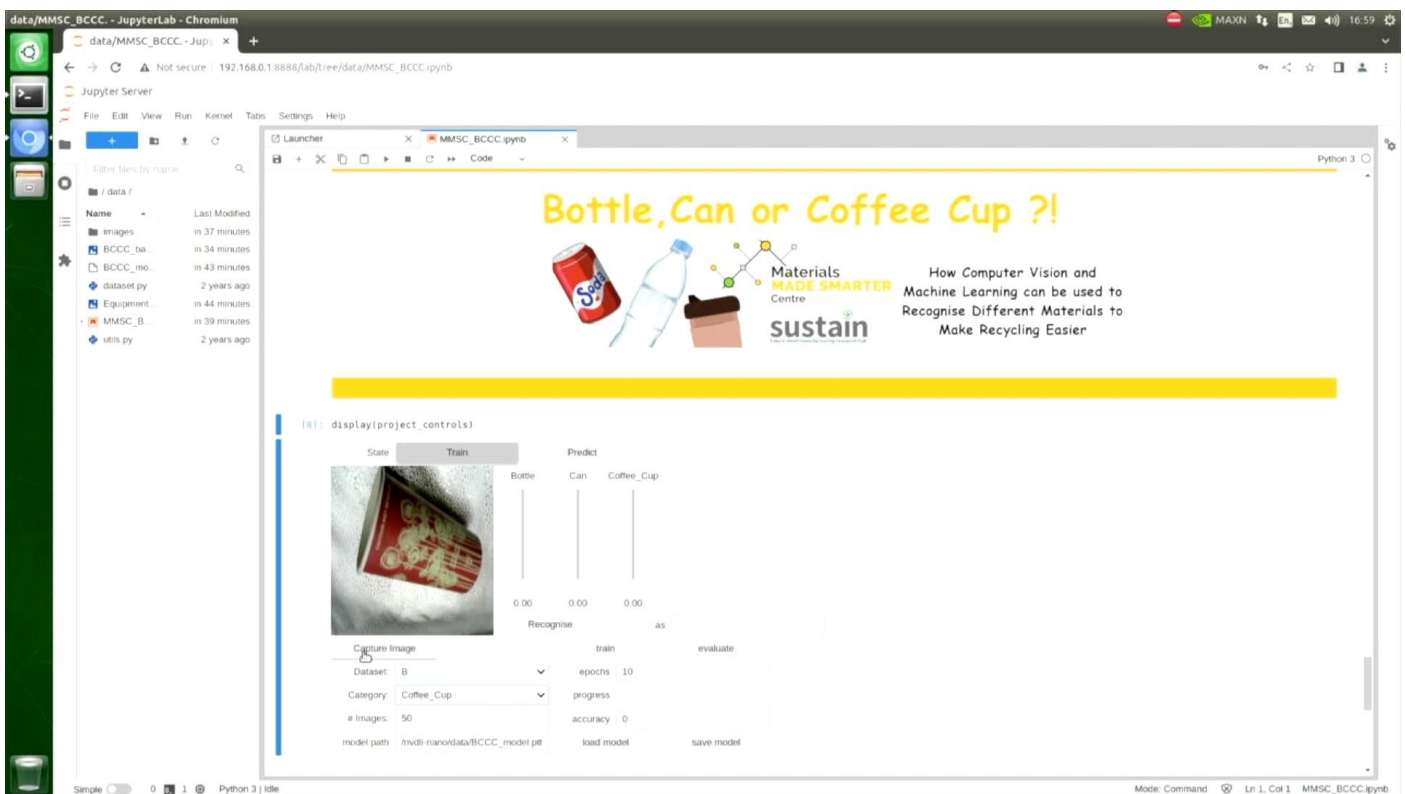
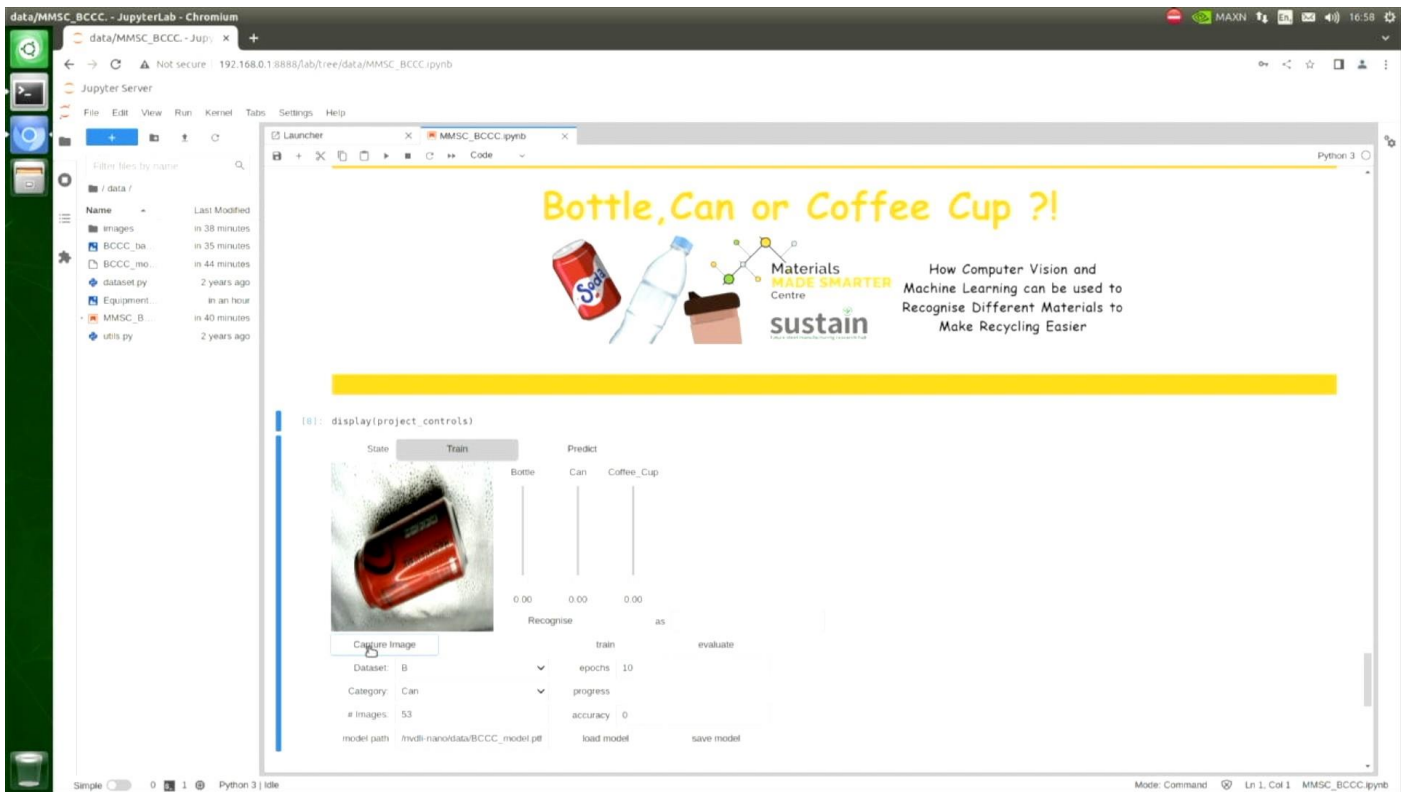


We are in training mode again and we need to select our new dataset B. We can see that there are already 50 images in each of the categories of that dataset as well.

We are going to add State images of the three new examples of each object. It will only need ten images of each to improve the system.

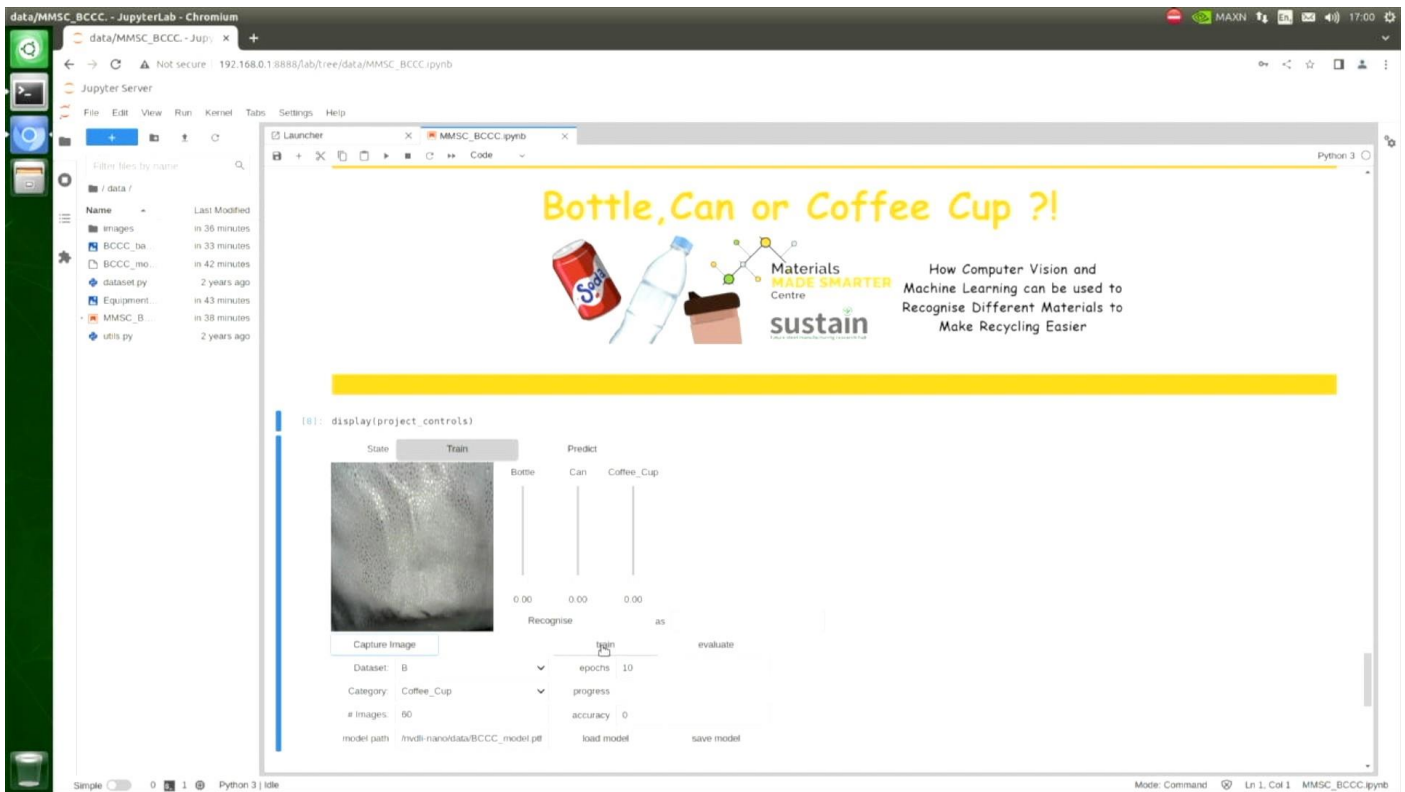


We place the new bottle in the system, click **Capture Image**, and then move the bottle to a different orientation and repeat 10 times.



We do the same for the new version of the can and the coffee cup. Make sure that your hand is not present in the captured image and that each object has a good selection of different orientations. Ten new images of each category will be sufficient so the dataset **BCCC\_B** contains 60 images of each category.

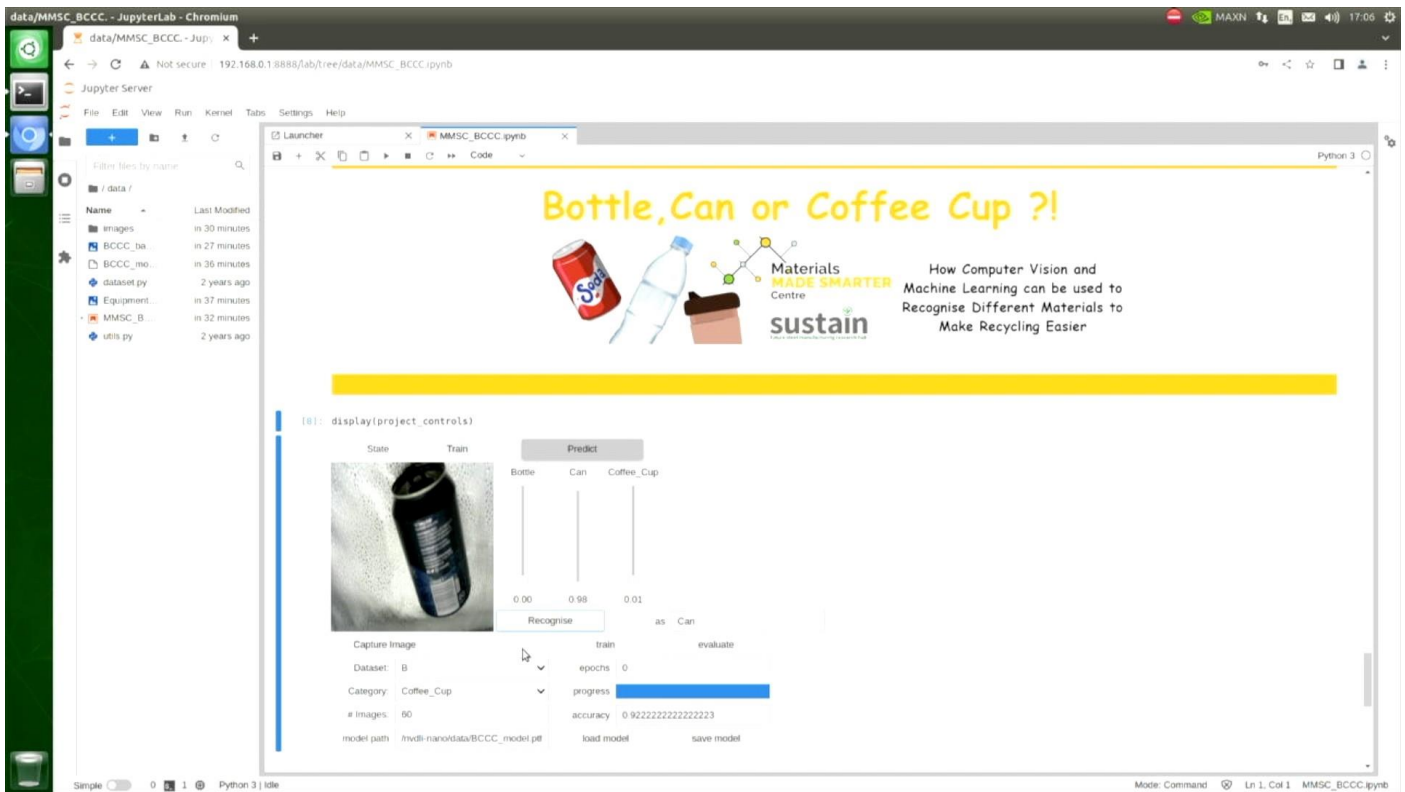
Thumbnails of the 60 images of the three objects, stored in **BCCC\_B** are attached to the end of this guide, 50 of the original objects and 10 of the new versions.



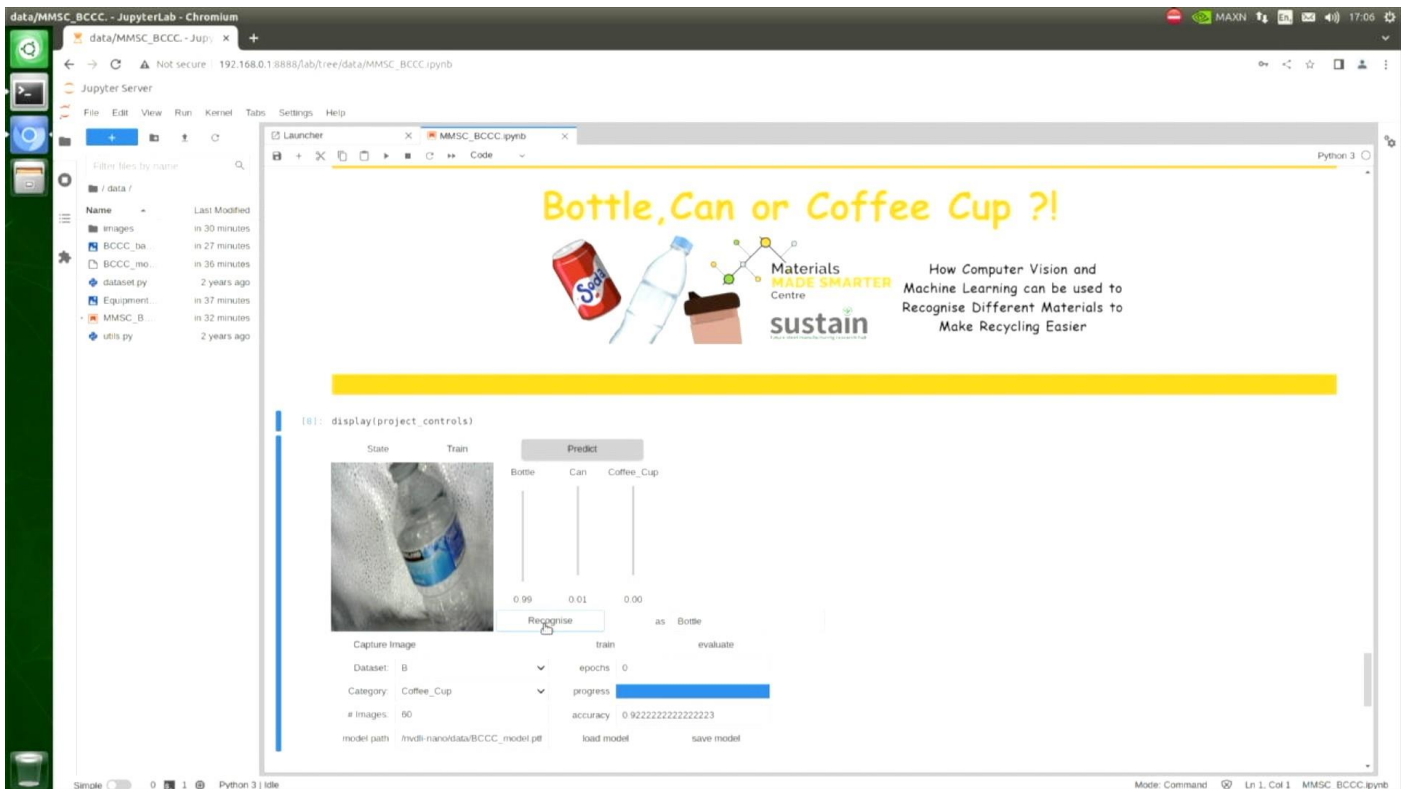
Again with 10 epochs, we are going to train the system on dataset B, 60 images, and see if just those 10 extra images helps it to recognise the coffee cup that it had trouble recognising before.

Even though it was able to recognise the bottle and the can without images of the new objects, it is important to have a **balanced dataset**, with an equal number of examples of each category.

It will take a bit longer than 4.6 minutes to train this time because there are 30 extra images, which with 10 epochs represents 300 extra training **iterations**.



After training, the first time the **Recognize** button is pressed, the process takes a little time as the model is loaded into memory, but after that the recognition process happens in real-time. The can is recognised with 98% confidence.



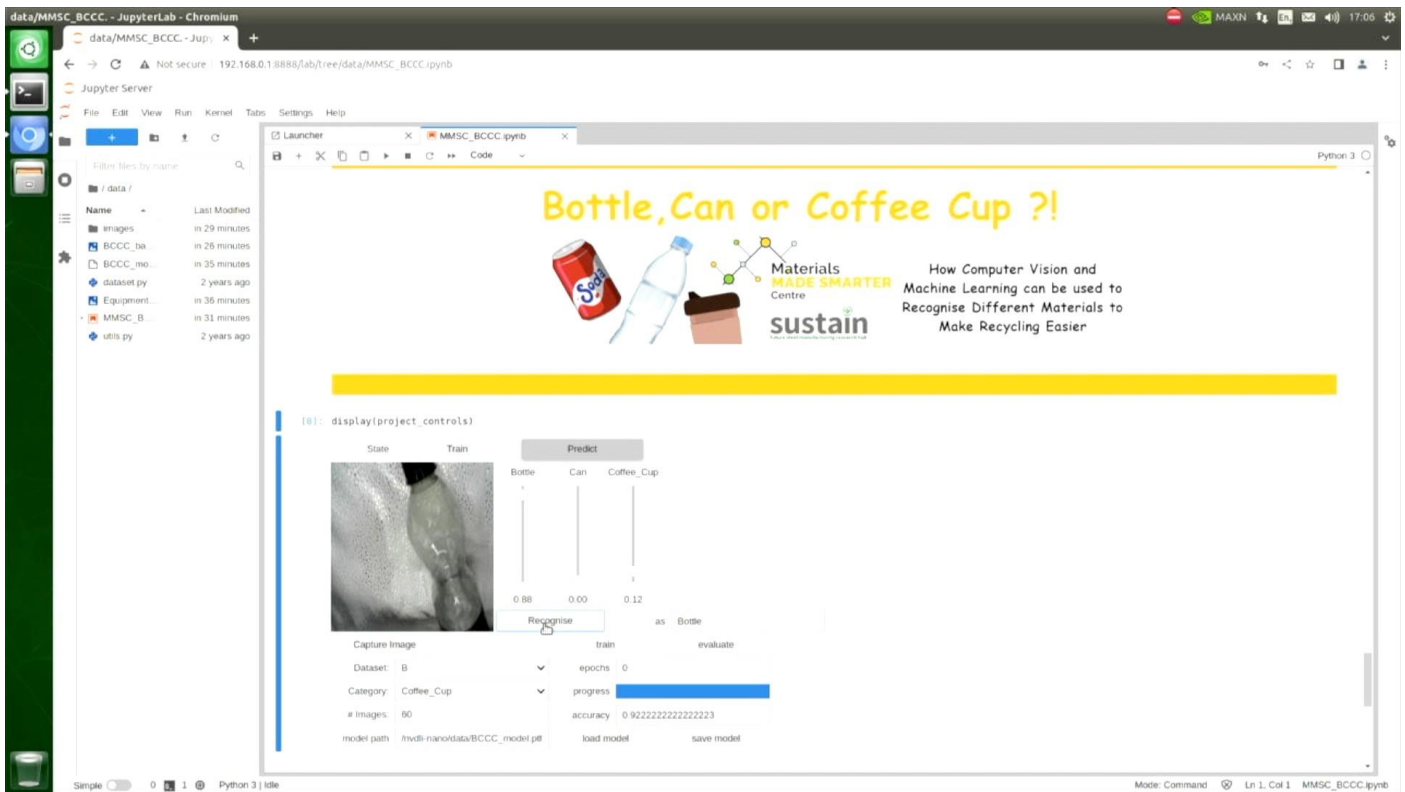
The original bottle with 99% confidence.

The screenshot shows a JupyterLab environment with a web application. The application has a header with the title "Bottle, Can or Coffee Cup ?!" and a logo for "Materials MADE SMARTER Centre sustain". Below the header, there is a section for "display(project\_controls)". This section contains a "State" image of a coffee cup, a "Train" button, and a "Predict" section with three radio buttons: "Bottle" (0.00), "Can" (0.00), and "Coffee\_Cup" (1.00). The "Recognise" button is active, and the prediction is "as Coffee\_Cup". Below the prediction, there are buttons for "train" and "evaluate". At the bottom, there are controls for "Dataset: B", "Category: Coffee\_Cup", "# Images: 60", and "model path: hvdb-nano\data\BCCC\_model.pkl". There are also buttons for "load model" and "save model". A progress bar is shown with an accuracy of 0.9222222222222223.

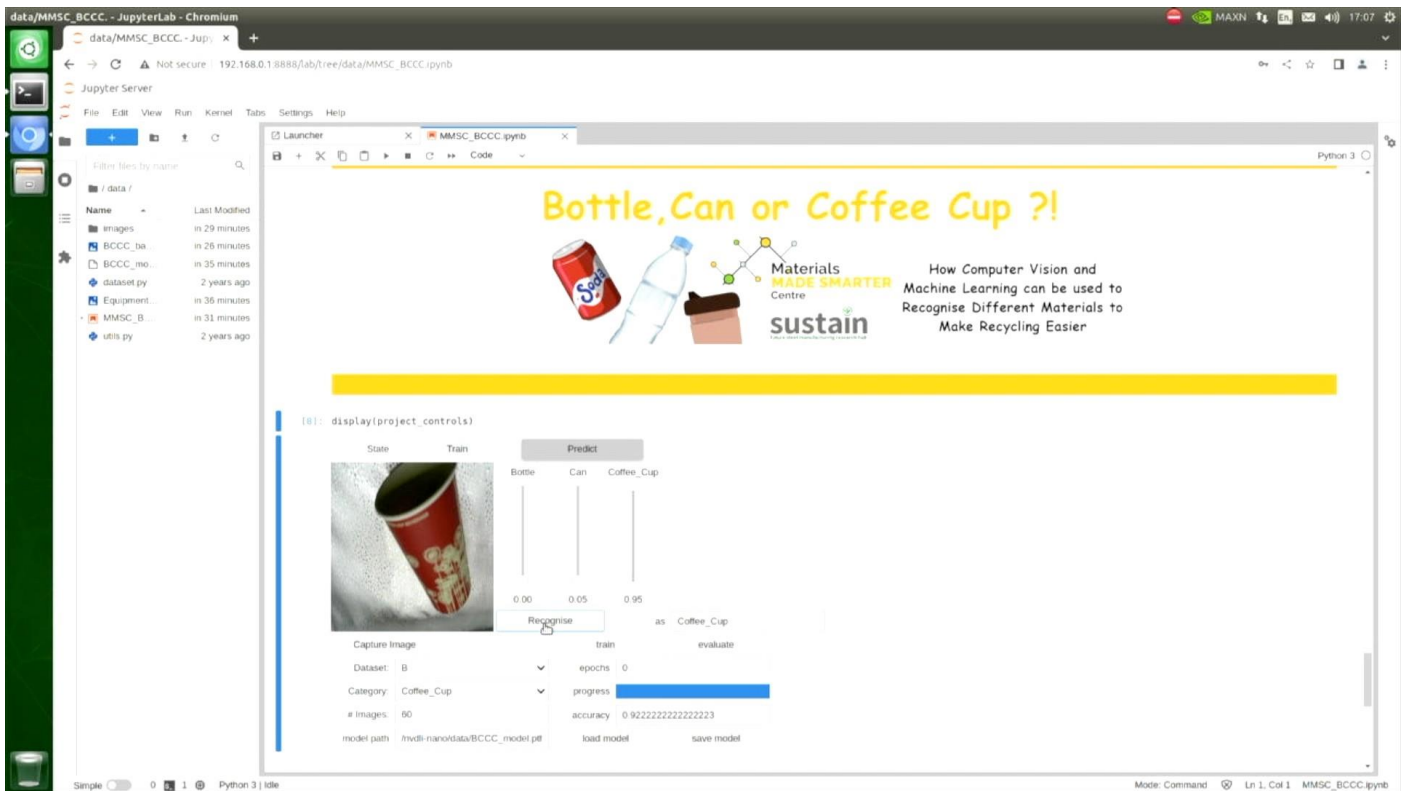
The original coffee cup with 100% confidence.

The screenshot shows the same JupyterLab environment and web application. The "State" image is now a red can. The "Train" button is active. The "Predict" section shows "Bottle" (0.00), "Can" (1.00), and "Coffee\_Cup" (0.00). The prediction is "as Can". The "Recognise" button is active, and the prediction is "as Can". Below the prediction, there are buttons for "train" and "evaluate". At the bottom, there are controls for "Dataset: B", "Category: Coffee\_Cup", "# Images: 60", and "model path: hvdb-nano\data\BCCC\_model.pkl". There are also buttons for "load model" and "save model". A progress bar is shown with an accuracy of 0.9222222222222223.

The new can with 100% confidence.



The new bottle with 88% confidence.



And then, finally, the new coffee cup with 95% confidence on the new coffee cup.

You can see in this straightforward demonstration how it is possible to re-train machine learning using just a few extra images to really improve its understanding of a new object without having to train from scratch every single time on exactly what is a bottle, a can or a coffee cup. Have fun playing around with the demonstration and I'll see you next time.



./data/images/BCCC\_A/Bottle/



0e51f56e-d430-11ef-9eca-48b02d9b1de8



1c4c3b2a-d430-11ef-9eca-48b02d9b1de8



02fd10f4-d440-11ef-ad7d-48b02d9b1de8



2c5d7042-d430-11ef-9eca-48b02d9b1de8



2f018d06-d430-11ef-9eca-48b02d9b1de8



3a23129a-d430-11ef-9eca-48b02d9b1de8



04c9bbbc-d430-11ef-9eca-48b02d9b1de8



09d10e9e-d430-11ef-9eca-48b02d9b1de8



13fe9ddc-d430-11ef-9eca-48b02d9b1de8



26a023d4-d430-11ef-9eca-48b02d9b1de8



32c15034-d430-11ef-9eca-48b02d9b1de8



093ed168-d440-11ef-ad7d-48b02d9b1de8



204fe1ae-d430-11ef-9eca-48b02d9b1de8



230ff56e-d430-11ef-9eca-48b02d9b1de8



1984a512-d430-11ef-9eca-48b02d9b1de8



06158388-d440-11ef-ad7d-48b02d9b1de8



a28e7276-d42f-11ef-9eca-48b02d9b1de8



a743f25a-d42f-11ef-9eca-48b02d9b1de8



abc6e0c6-d42f-11ef-9eca-48b02d9b1de8



aea09fee-d42f-11ef-9eca-48b02d9b1de8



b2a62cf8-d42f-11ef-9eca-48b02d9b1de8



b5a95e66-d42f-11ef-9eca-48b02d9b1de8



b89ad6c2-d42f-11ef-9eca-48b02d9b1de8



bc6d9f0a-d42f-11ef-9eca-48b02d9b1de8



bf7b19c-d42f-11ef-9eca-48b02d9b1de8



c2a84faa-d42f-11ef-9eca-48b02d9b1de8



c64aa216-d42f-11ef-9eca-48b02d9b1de8



c9770f8-d42f-11ef-9eca-48b02d9b1de8



cf254f1c-d42f-11ef-9eca-48b02d9b1de8



dfef111a-d42f-11ef-9eca-48b02d9b1de8



d84b2ad0-d42f-11ef-9eca-48b02d9b1de8



d5215262-d42f-11ef-9eca-48b02d9b1de8



dee0b57c-d42f-11ef-9eca-48b02d9b1de8



ef1e57ef6-d42f-11ef-9eca-48b02d9b1de8



ef62c2ea-d43f-11ef-ad7d-48b02d9b1de8



e9dbb502-d43f-11ef-ad7d-48b02d9b1de8



e41eb1b4-d43f-11ef-ad7d-48b02d9b1de8



e0065622-d43f-11ef-ad7d-48b02d9b1de8



eb120a44-d42f-11ef-9eca-48b02d9b1de8



eda8a2ee-d43f-11ef-ad7d-48b02d9b1de8



f3a066c4-d42f-11ef-9eca-48b02d9b1de8



f07c9002-d43f-11ef-ad7d-48b02d9b1de8



f9a698a-d43f-11ef-ad7d-48b02d9b1de8



f6297ab6-d42f-11ef-9eca-48b02d9b1de8



f7441c34-d43f-11ef-ad7d-48b02d9b1de8



f39829ae-d43f-11ef-ad7d-48b02d9b1de8



f0782252-d42f-11ef-9eca-48b02d9b1de8



fc6b86ea-d43f-11ef-ad7d-48b02d9b1de8



ff4e5b34-d42f-11ef-9eca-48b02d9b1de8



ffad70a2-d43f-11ef-ad7d-48b02d9b1de8

./data/images/BCCC\_A/Can/



065415e2-d43d-11ef-be21-48b02d9b1de8



072e6a8-d43d-11ef-be21-48b02d9b1de8



01c8c248-d43d-11ef-be21-48b02d9b1de8



2c6eee32-d43d-11ef-be21-48b02d9b1de8



2f3b470a-d43d-11ef-be21-48b02d9b1de8



3b68345c-d43d-11ef-be21-48b02d9b1de8



04ab0c3c-d43d-11ef-be21-48b02d9b1de8



5bef972e-d43d-11ef-be21-48b02d9b1de8



5f07618a-d43d-11ef-be21-48b02d9b1de8



6a0b0604-d43d-11ef-be21-48b02d9b1de8



6ecf3e62-d43d-11ef-be21-48b02d9b1de8



8b605480-d43d-11ef-be21-48b02d9b1de8



9b0cf674-d43d-11ef-be21-48b02d9b1de8



9fd4058-d43d-11ef-be21-48b02d9b1de8



40be8c12-d43d-11ef-be21-48b02d9b1de8



55d277ee-d43d-11ef-be21-48b02d9b1de8



57e16e00-d43d-11ef-be21-48b02d9b1de8



083e0afc-d43d-11ef-be21-48b02d9b1de8



97d8b6c8-d43d-11ef-be21-48b02d9b1de8



201d404-d43d-11ef-be21-48b02d9b1de8



649e032e-d43d-11ef-be21-48b02d9b1de8



790e7e88-d43d-11ef-be21-48b02d9b1de8



810a2920-d43d-11ef-be21-48b02d9b1de8



920a4494-d43d-11ef-be21-48b02d9b1de8



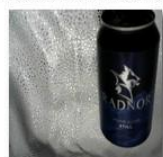
3322e78-d43d-11ef-be21-48b02d9b1de8



842383b8-d43d-11ef-be21-48b02d9b1de8



9555966c-d43d-11ef-be21-48b02d9b1de8



48044674-d43d-11ef-be21-48b02d9b1de8



a5be7ffa-d43d-11ef-be21-48b02d9b1de8



a8f34d46-d43d-11ef-be21-48b02d9b1de8



b0b371fe-d43d-11ef-be21-48b02d9b1de8



b4f1fa46-d43d-11ef-be21-48b02d9b1de8



b82d743e-d43d-11ef-be21-48b02d9b1de8



baee04c2-d43d-11ef-be21-48b02d9b1de8



c9a51816-d43d-11ef-be21-48b02d9b1de8



ce0a8440-d43d-11ef-be21-48b02d9b1de8



d3deb40-d43d-11ef-be21-48b02d9b1de8



d031cfee-d43d-11ef-be21-48b02d9b1de8



d882d410-d43d-11ef-be21-48b02d9b1de8



e01f1ab0-d43d-11ef-be21-48b02d9b1de8



e5f4f310-d43d-11ef-be21-48b02d9b1de8



e7fd2ef8-d43c-11ef-be21-48b02d9b1de8



e1451dc8-d43c-11ef-be21-48b02d9b1de8



eb9d30f8-d43c-11ef-be21-48b02d9b1de8



eed54c60-d43c-11ef-be21-48b02d9b1de8



f8a1fba2-d43d-11ef-be21-48b02d9b1de8



f14f893c-d43d-11ef-be21-48b02d9b1de8



f400703e-d43c-11ef-be21-48b02d9b1de8



f4572306-d43d-11ef-be21-48b02d9b1de8



fe95d0de-d43c-11ef-be21-48b02d9b1de8

./data/images/BCCC\_A/Coffee\_Cup/



0d828bc4-d432-11ef-9eca-48b02d9b1de8



0dd42eda-d431-11ef-9eca-48b02d9b1de8



6cb1fc9e-d430-11ef-9eca-48b02d9b1de8



6ee545a2-d430-11ef-9eca-48b02d9b1de8



7b5762fc-d430-11ef-9eca-48b02d9b1de8



7dc6bae2-d430-11ef-9eca-48b02d9b1de8



06fac0a8-d432-11ef-9eca-48b02d9b1de8



9fc8b146-d430-11ef-9eca-48b02d9b1de8



11de4830-d431-11ef-9eca-48b02d9b1de8



63d46b8e-d430-11ef-9eca-48b02d9b1de8



69eb2058-d430-11ef-9eca-48b02d9b1de8



837a95b2-d430-11ef-9eca-48b02d9b1de8



865ef0c0-d430-11ef-9eca-48b02d9b1de8



896e4d42-d430-11ef-9eca-48b02d9b1de8



915c126e-d430-11ef-9eca-48b02d9b1de8



7493a912-d430-11ef-9eca-48b02d9b1de8



8057ee20-d430-11ef-9eca-48b02d9b1de8



09629866-d431-11ef-9eca-48b02d9b1de8



040125c4-d432-11ef-9eca-48b02d9b1de8



0568855c-d431-11ef-9eca-48b02d9b1de8



15951256-d431-11ef-9eca-48b02d9b1de8



71634428-d430-11ef-9eca-48b02d9b1de8



77759992-d430-11ef-9eca-48b02d9b1de8



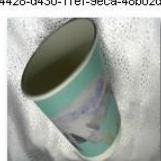
a91a6c70-d430-11ef-9eca-48b02d9b1de8



a5338bc0-d430-11ef-9eca-48b02d9b1de8



aec1c448-d430-11ef-9eca-48b02d9b1de8



b282b4ac-d430-11ef-9eca-48b02d9b1de8



b8023880-d430-11ef-9eca-48b02d9b1de8



bbf5350a-d430-11ef-9eca-48b02d9b1de8



bf2ca352-d430-11ef-9eca-48b02d9b1de8



c4a7661a-d430-11ef-9eca-48b02d9b1de8



c8e33d48-d430-11ef-9eca-48b02d9b1de8



ce945baa-d430-11ef-9eca-48b02d9b1de8



d4c72c40-d431-11ef-9eca-48b02d9b1de8



d6cf1f3c-d431-11ef-9eca-48b02d9b1de8



d935a820-d430-11ef-9eca-48b02d9b1de8



dc86411a-d430-11ef-9eca-48b02d9b1de8



dff377fe-d431-11ef-9eca-48b02d9b1de8



e2da3436-d430-11ef-9eca-48b02d9b1de8



e325d4c6-d431-11ef-9eca-48b02d9b1de8



e862bf6c-d431-11ef-9eca-48b02d9b1de8



ebf46a7c-d431-11ef-9eca-48b02d9b1de8



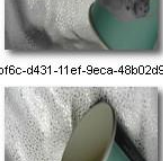
ec77b93c-d430-11ef-9eca-48b02d9b1de8



eeb401fa-d431-11ef-9eca-48b02d9b1de8



f21c4906-d431-11ef-9eca-48b02d9b1de8



f813a7a6-d430-11ef-9eca-48b02d9b1de8



f881c114-d431-11ef-9eca-48b02d9b1de8



f676817a-d430-11ef-9eca-48b02d9b1de8



fbf4aec8-d431-11ef-9eca-48b02d9b1de8



fc978a42-d430-11ef-9eca-48b02d9b1de8

./data/images/BCCC\_B/Bottle/



0e51f56e-d430-11ef-9eca-48b02d9b1de8



1c4c3b2a-d430-11ef-9eca-48b02d9b1de8



02fd10f4-d440-11ef-ad7d-48b02d9b1de8



2c5d7042-d430-11ef-9eca-48b02d9b1de8



2f018d06-d430-11ef-9eca-48b02d9b1de8



3a23129a-d430-11ef-9eca-48b02d9b1de8



04c9bbbc-d430-11ef-9eca-48b02d9b1de8



09d10e9e-d430-11ef-9eca-48b02d9b1de8



13fe9ddc-d430-11ef-9eca-48b02d9b1de8



26a023d4-d430-11ef-9eca-48b02d9b1de8



32c15034-d430-11ef-9eca-48b02d9b1de8



093ed168-d440-11ef-ad7d-48b02d9b1de8



204fe1ae-d430-11ef-9eca-48b02d9b1de8



230ff56e-d430-11ef-9eca-48b02d9b1de8



1984a512-d430-11ef-9eca-48b02d9b1de8



06158388-d440-11ef-ad7d-48b02d9b1de8



a28e7276-d42f-11ef-9eca-48b02d9b1de8



a743f25a-d42f-11ef-9eca-48b02d9b1de8



abc6e0c6-d42f-11ef-9eca-48b02d9b1de8



aea09fee-d42f-11ef-9eca-48b02d9b1de8



b2a62c16-d42f-11ef-9eca-48b02d9b1de8



b5a95e66-d42f-11ef-9eca-48b02d9b1de8



b89ad6c2-d42f-11ef-9eca-48b02d9b1de8



bc6d9f0a-d42f-11ef-9eca-48b02d9b1de8



bf7b19c-d42f-11ef-9eca-48b02d9b1de8



c2a841aa-d42f-11ef-9eca-48b02d9b1de8



c64aa216-d42f-11ef-9eca-48b02d9b1de8



c9770f8-d42f-11ef-9eca-48b02d9b1de8



cf254f1c-d42f-11ef-9eca-48b02d9b1de8



dfef111a-d42f-11ef-9eca-48b02d9b1de8



d84b2ad0-d42f-11ef-9eca-48b02d9b1de8



d5215262-d42f-11ef-9eca-48b02d9b1de8



dee0b57c-d42f-11ef-9eca-48b02d9b1de8



ef1e57ef6-d42f-11ef-9eca-48b02d9b1de8



ef62c2ea-d43f-11ef-ad7d-48b02d9b1de8



e9dbb502-d43f-11ef-ad7d-48b02d9b1de8



e41eb1b4-d43f-11ef-ad7d-48b02d9b1de8



e0065622-d43f-11ef-ad7d-48b02d9b1de8



eb120a44-d42f-11ef-9eca-48b02d9b1de8



eda8a2ee-d43f-11ef-ad7d-48b02d9b1de8



f3a066c4-d42f-11ef-9eca-48b02d9b1de8



f07c9002-d43f-11ef-ad7d-48b02d9b1de8



f9a698a-d43f-11ef-ad7d-48b02d9b1de8



f6297ab6-d42f-11ef-9eca-48b02d9b1de8



f7441c34-d43f-11ef-ad7d-48b02d9b1de8



f39629ae-d43f-11ef-ad7d-48b02d9b1de8



f0782252-d42f-11ef-9eca-48b02d9b1de8



fc6b86ea-d43f-11ef-ad7d-48b02d9b1de8



ff4e5b34-d42f-11ef-9eca-48b02d9b1de8



ffad70a2-d43f-11ef-ad7d-48b02d9b1de8



bd44e426-d430-11ef-bab2-48b02d9b1de8



c625b1e4-d430-11ef-bab2-48b02d9b1de8



c1439dd0-d430-11ef-bab2-48b02d9b1de8



c8913f7a-d430-11ef-bab2-48b02d9b1de8



c3509664-d430-11ef-bab2-48b02d9b1de8



cb54d974-d430-11ef-bab2-48b02d9b1de8



ce7173a6-d430-11ef-bab2-48b02d9b1de8



d0b16b44-d430-11ef-bab2-48b02d9b1de8



d71d6b40-d430-11ef-bab2-48b02d9b1de8



d427c80e-d430-11ef-bab2-48b02d9b1de8

./data/images/BCCC\_B/Can/



065415e2-d43d-11ef-be21-48b02d9b1de8



072e6a8-d43d-11ef-be21-48b02d9b1de8



01c8c248-d43d-11ef-be21-48b02d9b1de8



2c6eee32-d43d-11ef-be21-48b02d9b1de8



2f3b470a-d43d-11ef-be21-48b02d9b1de8



3b68345c-d43d-11ef-be21-48b02d9b1de8



04ab0c3c-d43d-11ef-be21-48b02d9b1de8



5bef972e-d43d-11ef-be21-48b02d9b1de8



5f07618a-d43d-11ef-be21-48b02d9b1de8



6a0b0604-d43d-11ef-be21-48b02d9b1de8



6ecf3e62-d43d-11ef-be21-48b02d9b1de8



8b605480-d43d-11ef-be21-48b02d9b1de8



9b0cf674-d43d-11ef-be21-48b02d9b1de8



9fd4058-d43d-11ef-be21-48b02d9b1de8



40be8c12-d43d-11ef-be21-48b02d9b1de8



55d277ee-d43d-11ef-be21-48b02d9b1de8



57e16e00-d43d-11ef-be21-48b02d9b1de8



083e0afc-d43d-11ef-be21-48b02d9b1de8



97d8b6c8-d43d-11ef-be21-48b02d9b1de8



201d404-d43d-11ef-be21-48b02d9b1de8



649e032e-d43d-11ef-be21-48b02d9b1de8



790e7e88-d43d-11ef-be21-48b02d9b1de8



810a2920-d43d-11ef-be21-48b02d9b1de8



920a4494-d43d-11ef-be21-48b02d9b1de8



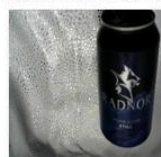
3322e78-d43d-11ef-be21-48b02d9b1de8



842383b8-d43d-11ef-be21-48b02d9b1de8



9555966c-d43d-11ef-be21-48b02d9b1de8



48044674-d43d-11ef-be21-48b02d9b1de8



a5be7ffa-d43d-11ef-be21-48b02d9b1de8



a8f34d46-d43d-11ef-be21-48b02d9b1de8



b0b371fe-d43d-11ef-be21-48b02d9b1de8



b4f1fa46-d43d-11ef-be21-48b02d9b1de8



b82d743e-d43d-11ef-be21-48b02d9b1de8



baee04c2-d43d-11ef-be21-48b02d9b1de8



c9a51816-d43d-11ef-be21-48b02d9b1de8



ce0a8440-d43d-11ef-be21-48b02d9b1de8



d3deb40-d43d-11ef-be21-48b02d9b1de8



d031cfee-d43d-11ef-be21-48b02d9b1de8



d882d410-d43d-11ef-be21-48b02d9b1de8



e01f1ab0-d43d-11ef-be21-48b02d9b1de8



e5f4f310-d43d-11ef-be21-48b02d9b1de8



e7fd2ef8-d43c-11ef-be21-48b02d9b1de8



e1451dc8-d43c-11ef-be21-48b02d9b1de8



eb9d30f8-d43c-11ef-be21-48b02d9b1de8



eed54c60-d43c-11ef-be21-48b02d9b1de8



f8a1fba2-d43d-11ef-be21-48b02d9b1de8



f14f893c-d43d-11ef-be21-48b02d9b1de8



f400703e-d43c-11ef-be21-48b02d9b1de8



f4572306-d43d-11ef-be21-48b02d9b1de8



fe95d0de-d43c-11ef-be21-48b02d9b1de8



00d885e6-d431-11ef-bab2-48b02d9b1de8



03ee3dc0-d431-11ef-bab2-48b02d9b1de8



e5a4c226-d430-11ef-bab2-48b02d9b1de8



e19baaa0-d430-11ef-bab2-48b02d9b1de8



e92a376e-d430-11ef-bab2-48b02d9b1de8



ecdaf97a-d430-11ef-bab2-48b02d9b1de8



f9cfb8a0-d430-11ef-bab2-48b02d9b1de8



f14cd3f2-d430-11ef-bab2-48b02d9b1de8



f44e553a-d430-11ef-bab2-48b02d9b1de8



fd3fa140-d430-11ef-bab2-48b02d9b1de8

./data/images/BCCC\_B/Coffee\_Cup/



0d828bc4-d432-11ef-9eca-48b02d9b1de8



0dd42eda-d431-11ef-9eca-48b02d9b1de8



6cb1fc9e-d430-11ef-9eca-48b02d9b1de8



6ee545a2-d430-11ef-9eca-48b02d9b1de8



7b5762fc-d430-11ef-9eca-48b02d9b1de8



7dc6bae2-d430-11ef-9eca-48b02d9b1de8



06fac0a8-d432-11ef-9eca-48b02d9b1de8



9fc8b146-d430-11ef-9eca-48b02d9b1de8



11de4830-d431-11ef-9eca-48b02d9b1de8



63d46b8e-d430-11ef-9eca-48b02d9b1de8



69eb2058-d430-11ef-9eca-48b02d9b1de8



837a95b2-d430-11ef-9eca-48b02d9b1de8



865ef0c0-d430-11ef-9eca-48b02d9b1de8



896e4d42-d430-11ef-9eca-48b02d9b1de8



915c126e-d430-11ef-9eca-48b02d9b1de8



7493a912-d430-11ef-9eca-48b02d9b1de8



8057ee20-d430-11ef-9eca-48b02d9b1de8



0962986-d431-11ef-9eca-48b02d9b1de8



040125c4-d432-11ef-9eca-48b02d9b1de8



0568855c-d431-11ef-9eca-48b02d9b1de8



15951256-d431-11ef-9eca-48b02d9b1de8



71634428-d430-11ef-9eca-48b02d9b1de8



7775992-d430-11ef-9eca-48b02d9b1de8



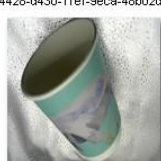
a91a6c70-d430-11ef-9eca-48b02d9b1de8



a5338bc0-d430-11ef-9eca-48b02d9b1de8



aec1c448-d430-11ef-9eca-48b02d9b1de8



b282b4ac-d430-11ef-9eca-48b02d9b1de8



b8023880-d430-11ef-9eca-48b02d9b1de8



bbf5350a-d430-11ef-9eca-48b02d9b1de8



bf2ca352-d430-11ef-9eca-48b02d9b1de8



c4a7661a-d430-11ef-9eca-48b02d9b1de8



c8e33d48-d430-11ef-9eca-48b02d9b1de8



ce945baa-d430-11ef-9eca-48b02d9b1de8



d4c72c40-d431-11ef-9eca-48b02d9b1de8



d6cf1f3c-d431-11ef-9eca-48b02d9b1de8



d935a820-d430-11ef-9eca-48b02d9b1de8



dc86411a-d430-11ef-9eca-48b02d9b1de8



dff377fe-d431-11ef-9eca-48b02d9b1de8



e2da3436-d430-11ef-9eca-48b02d9b1de8



e325d4c6-d431-11ef-9eca-48b02d9b1de8



e862bf6c-d431-11ef-9eca-48b02d9b1de8



ebf46a7c-d431-11ef-9eca-48b02d9b1de8



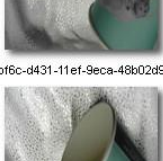
ec77b93c-d430-11ef-9eca-48b02d9b1de8



eeb401fa-d431-11ef-9eca-48b02d9b1de8



f21c4906-d431-11ef-9eca-48b02d9b1de8



f813a7a6-d430-11ef-9eca-48b02d9b1de8



f881c114-d431-11ef-9eca-48b02d9b1de8



f676817a-d430-11ef-9eca-48b02d9b1de8



fbf4aec8-d431-11ef-9eca-48b02d9b1de8



fc978a42-d430-11ef-9eca-48b02d9b1de8





0d6296b2-d431-11ef-bab2-48b02d9b1de8



1bae60a2-d431-11ef-bab2-48b02d9b1de8



1113c3a4-d431-11ef-bab2-48b02d9b1de8



2b956e64-d431-11ef-bab2-48b02d9b1de8



2f181aa2-d431-11ef-bab2-48b02d9b1de8



22d6dct6-d431-11ef-bab2-48b02d9b1de8



150d64aa-d431-11ef-bab2-48b02d9b1de8



286c2c48-d431-11ef-bab2-48b02d9b1de8



1208d064-d431-11ef-bab2-48b02d9b1de8



18066684-d431-11ef-bab2-48b02d9b1de8