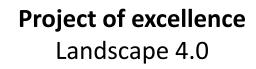
# POLITECNICO<br/>DI TORINOEuroDrying'20197thEuropean Drying Conference

## Recognition of inlet wet food into drying process through a deep learning approach Politecnico di Torino, Italy – July 10-12, 2019

**Roberto Moscetti<sup>a</sup>**, S. Massaro<sup>a</sup>, G. Chillemi<sup>a</sup>, N. Sanna<sup>a</sup>, B. Sturm<sup>b</sup>, S.S. Nallan Chakravatula<sup>a</sup>, R. Massantini<sup>a</sup> <sup>a</sup>Department for Innovation in Biological, Agro-food and Forest systems (DIBAF), University of Tuscia, Viterbo (Italy) <sup>b</sup>Department of Agricultural and Bisosystems Engineering, University of Kassel (Germany) <sup>c</sup> rmoscetti@unitus.it\*





ALIAN MINISTRY OF EDUCATION, UNIVERSITY AND RESEARCH

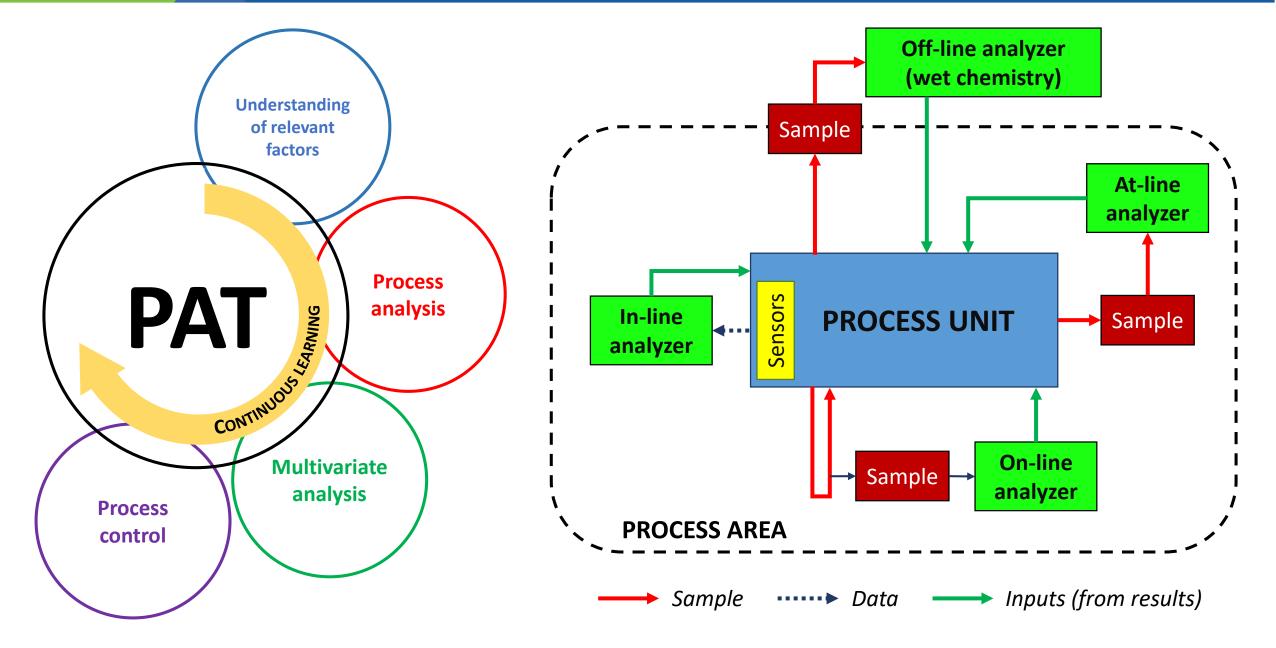








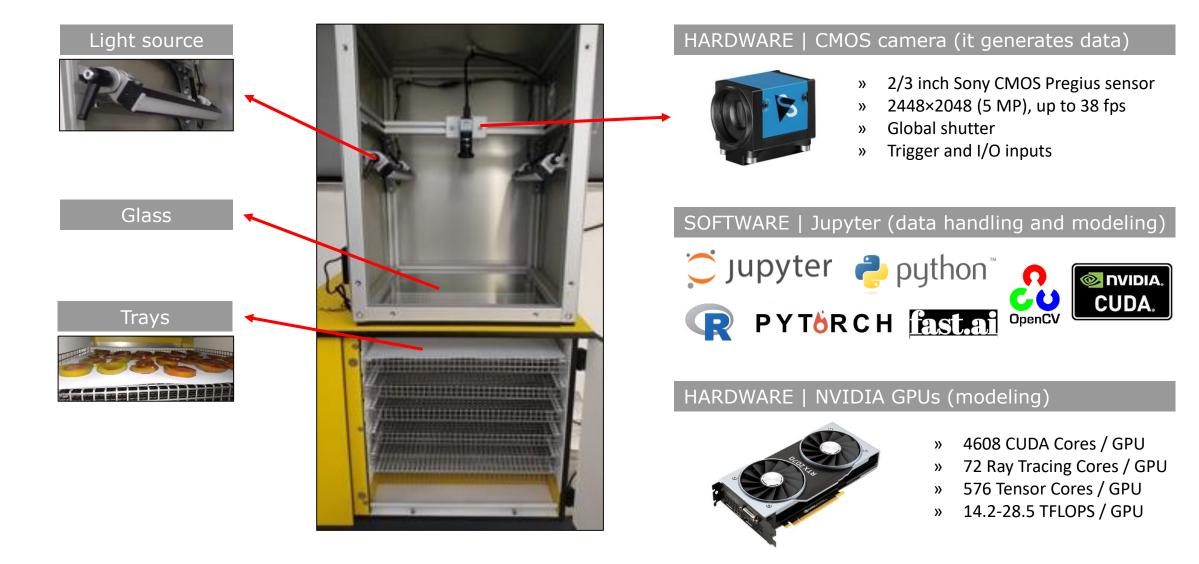








#### **» Our pilot dryer** | *Hardware and software*



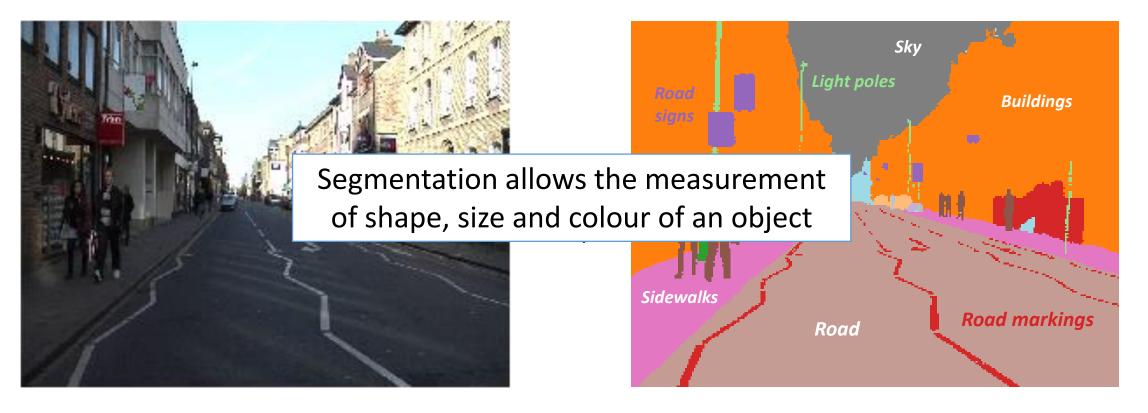


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### » What does CV applied to a dryer have to deal with? | The image segmentation problem...

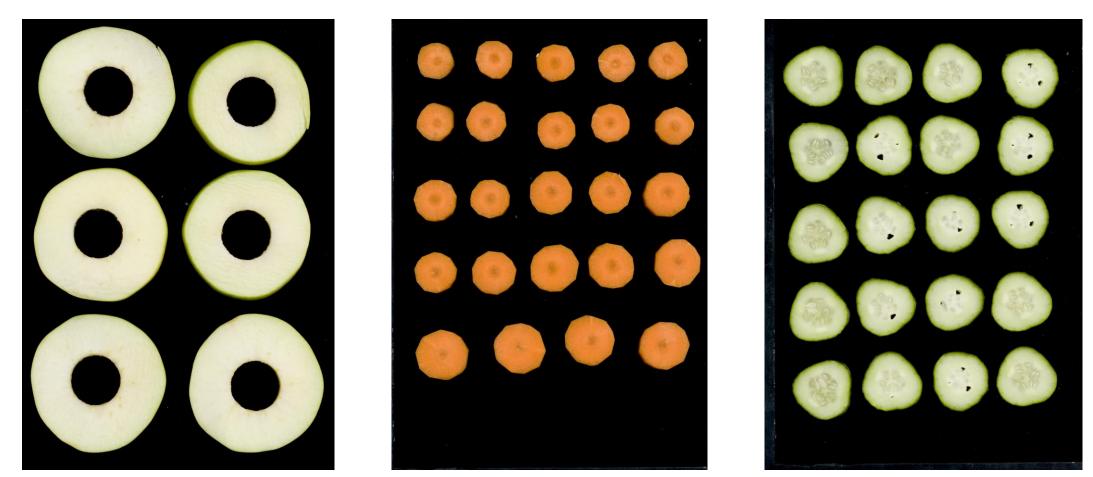
- > Classical segmentation consists in splitting an image into several coherent parts, without any attempt to understand what these parts represent
- Semantic segmentation attempts to partition the image into semantically meaningful parts, and to classify each part into one of the pre-determined classes





#### » What does CV applied to a dryer have to deal with? | The image recognition problem...

Recognize a product and set the proper process parameters (temperature, air flow and relative humidity)



Apple slices

Carrot slices

Cucumber slices

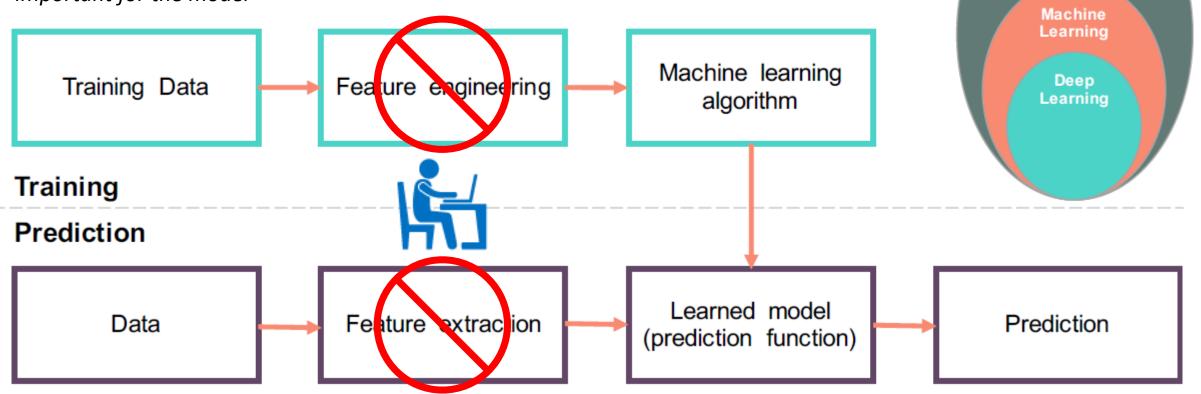
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#### **»** Artificial Intelligence | Machine learning vs deep learning

Machine learning requires feature engineering, while deep learning does not require it: algorithm automatically learns how to perform feature extraction and which features are important for the model



Modified from: Sorin Cheran's presentation titled 'About AI' (2018)

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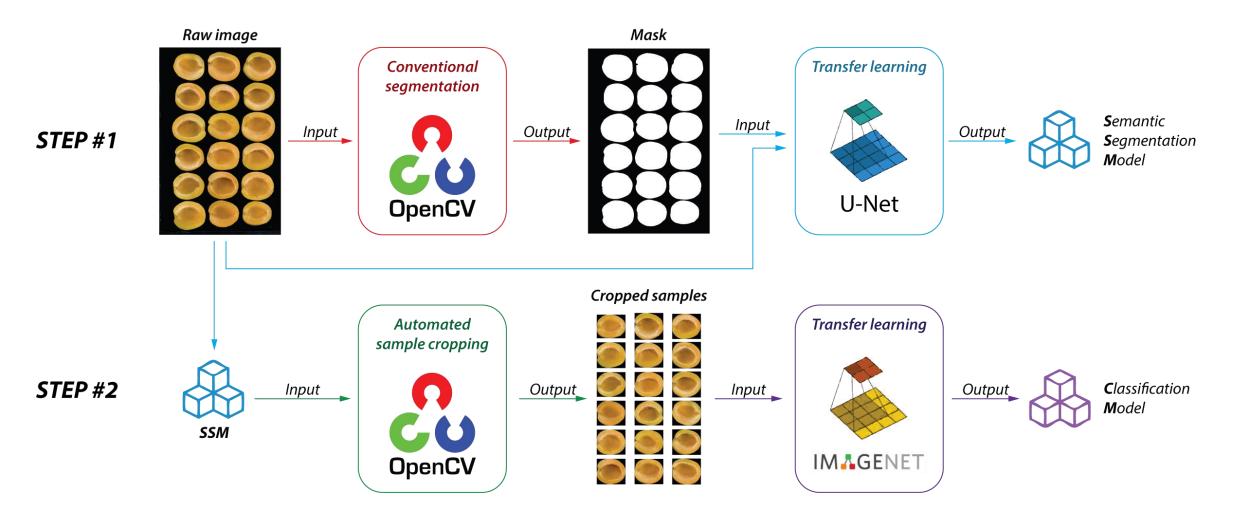
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Artificial Intelligence





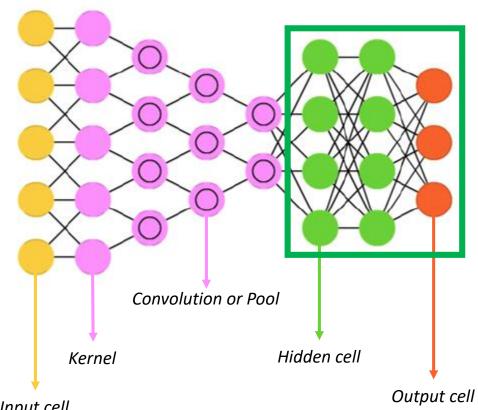
**» Our DNN approach** | *Semantic segmentation and Image classification* 





#### **How to make the model training much easier** | *the transfer learning approach* **>>**

Transfer learning is a technique that shortcuts much of this by taking a piece of a model that has already been trained on a related task and reusing it in a new model (source: Google Tensorflow website, 2019)



Last layers are the only ones that are retrained

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Input cell



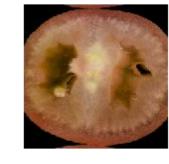


#### **» The dataset** | *100+ images per class of product*

- 1. Apricot
- 2. Banana
- 3. Carrot
- 4. Cucumber
- 5. Champignon (or white button, mushroom)
- 6. Cherry
- 7. Onion
- 8. Kiwifruit
- 9. Lime
- 10. Apple
- 11. Potato
- 12. Chilli pepper
- 13. Pear
- 14. Peach
- 15. Red plum
- 16. Zucchini
- 17. Cherry tomato
- 18. San Marzano tomato

Calibration set 70% - Prediction set 30%



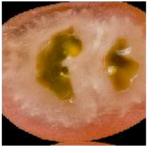














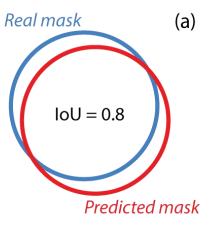


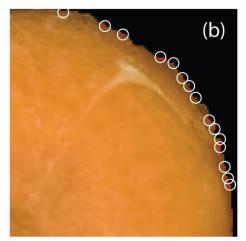
CNN models were retrained by perfoming data augmentation



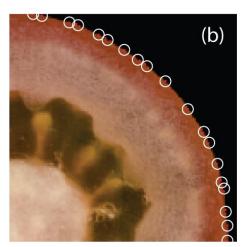


#### » **Results** | Semantic segmentation



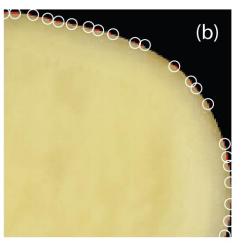


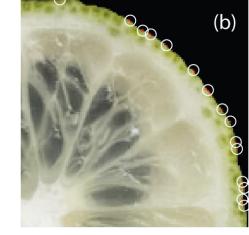
Carrot slice

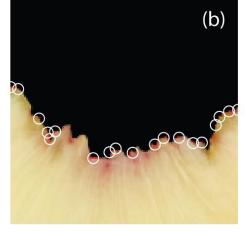


Plum tomato slice

Intersection over Union IoU > 99%







Potato slice

Lime slice

Peach slice

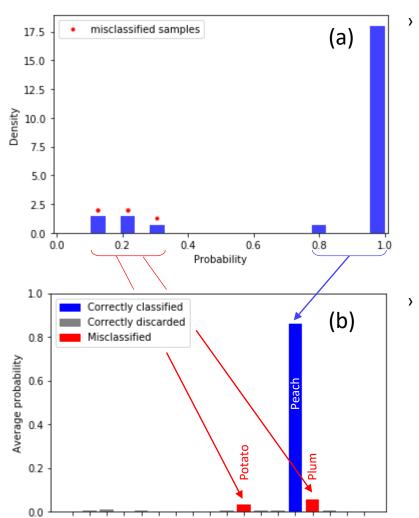
Example of IoU (a) and images of samples with misclassified pixels (b)



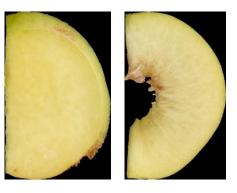


#### **» Results** | *Product recognition*

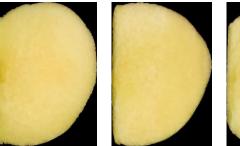
Model performance Accuracy = 0.992



Peach slices misclassified as red plum slices



Peach slices misclassified as potato slices









**» Results** | Training performance: comparison between CPU and GPU

Model	Learning	Epochs	Batch	Runtime	Training time
	rate		size	system	(hh:mm:ss)
СМ	1E-03	3	64	CPU	00:07:47
				GPU	00:00:26
SSM	5E-05	10	8	CPU	06:44:03
				GPU	00:22:31

CM: classification model SSM: semantic segmentation model

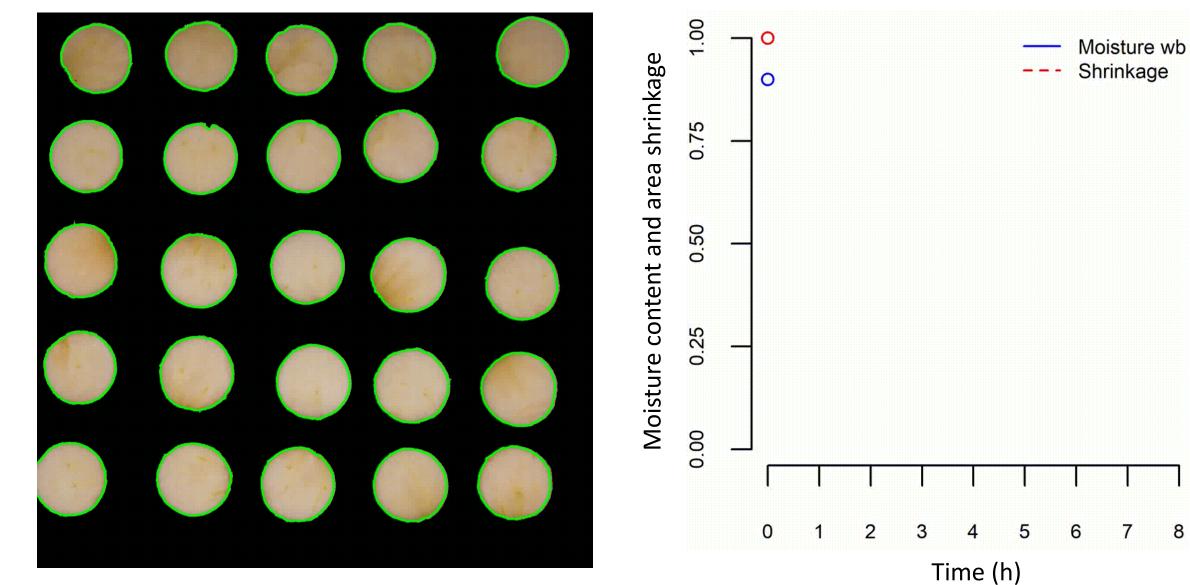
CPU mod. Ryzen 5 1400 (AMD Inc. CA, USA) GPU mod. RTX 2070 8 GB (NVIDIA Corp. CA, USA).





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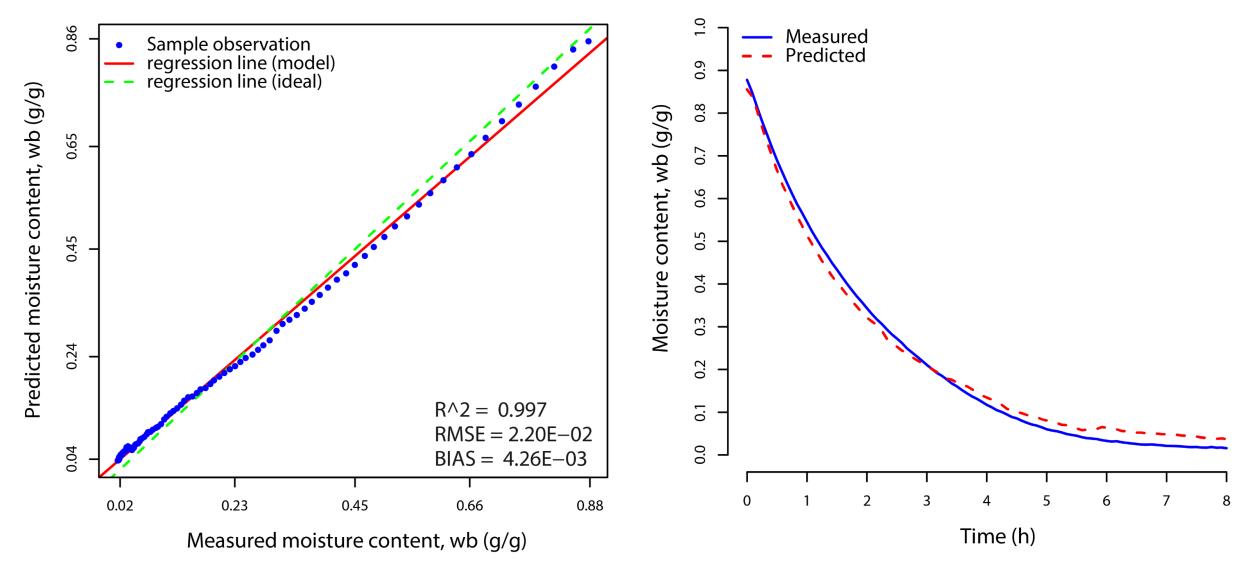
#### **Results** | Shape and size measurements »







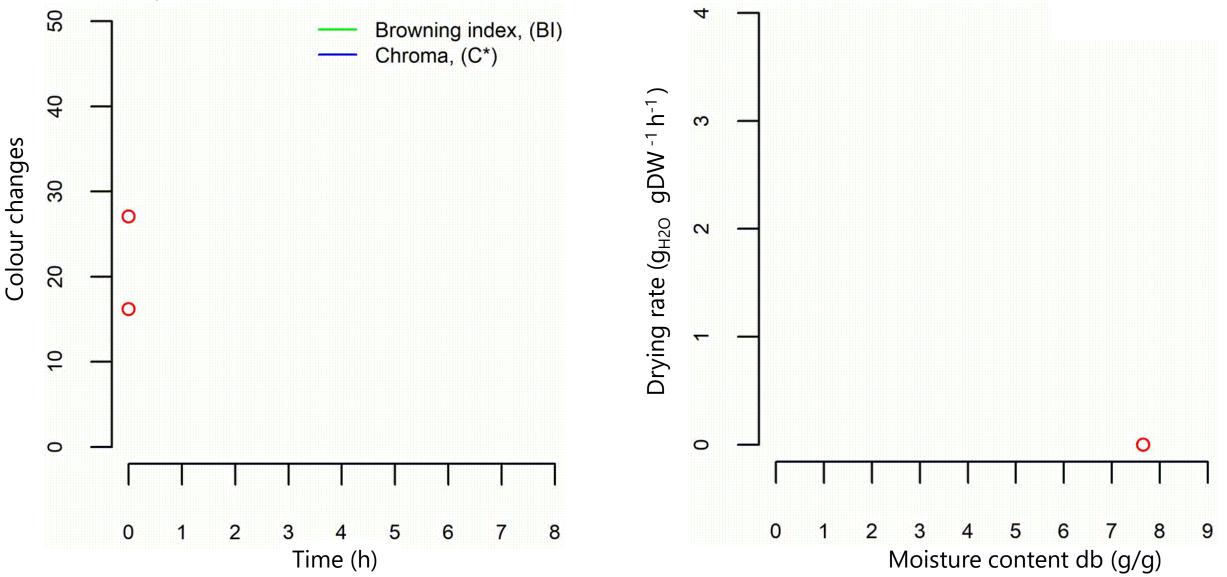
#### **» Results** | *Prediction of changes in moisture content*













#### **» Next step?** | *The implementation of CNNs in a drier*

## JETSON NANO DEVKIT SPECS

	PROCESSOR		INTERFACES
CPU 64-bit C	64-bit Quad-core ARM A57 @ 1.43GHz	USB	(4x) USB 3.0 A (Host)   USB 2.0 Micro B (Device)
		Camera	MIPI CSI-2 x2 (15-position Flex Connector)
PU 128-core M	128-core NVIDIA Maxwell @ 921MHz	Display	HDMI   DisplayPort
		Networking	Gigabit Ethernet (RJ45, PoE)
ory	4GB 64-bit LPDDR4 @ 1600MHz   25.6GB/s	Wireless	M.2 Key-E with PCIe x1
ideo Encoder 4Kp30		Storage	MicroSD card (16GB UHS-1 recommended minimum
	4Kp30   (4x) 1080p30   (2x) 1080p60	40-Pin Header	UART   SPI   I2C   I2S   Audio Clock   GPIOs
		Power	5V DC (µUSB, Barrel Jack, PoE) - 5W   10W
o Decoder	4Kp60   (2x) 4Kp30   (8x) 1080p30   (4x) 1080p60	Size	80x100mm

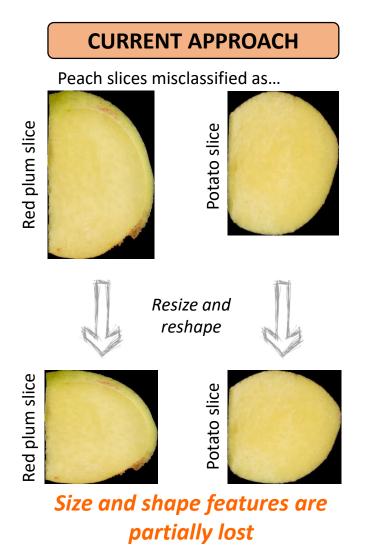
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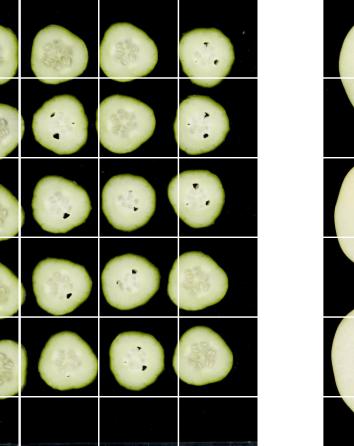


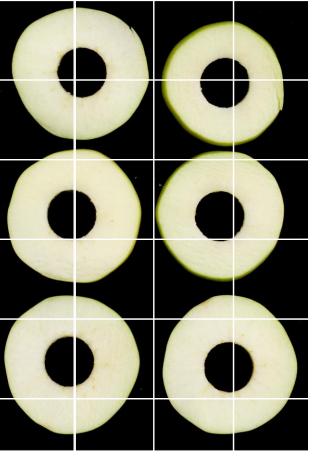


**» Next step?** | *Make the CNN models sensitive to differences in products size and shape* 



**NEW APPROACH** 



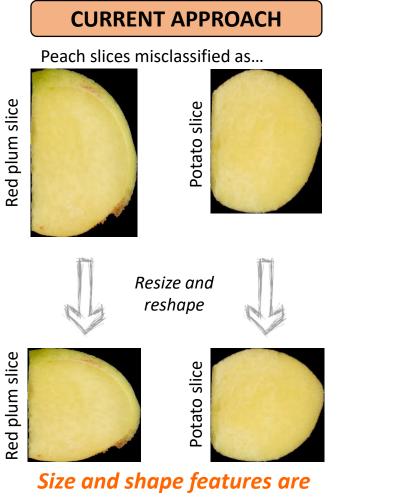


\* Peach slices misclassified as...





**» Next step?** | Make the CNN models sensitive to differences in products size and shape



partially lost

#### NEW APPROACH



Size and shape features are fully available



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# **THANK YOU FOR YOUR ATTENTION**