



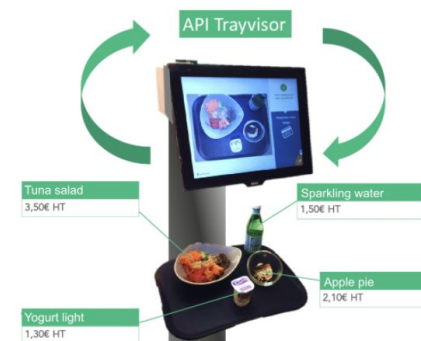
Deep learning based food instance Recognition

Classification - Segmentation Deep Learning Model

Use case: Self Service - Virtual Cashier - Calories Estimation



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Moscow - Melbourne - Paris
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- Introduction
- Methodology
- Result/ Discussion
- Conclusion



Introduction

- **Goal** : Experience real-life AI project from (A-Z) scratch
- According to WHO 20% of deaths worldwide are attributable to an unhealthy diet nowadays.
- 39% of adults aged 18 years and over were overweight in 2016, and 13% were obese.



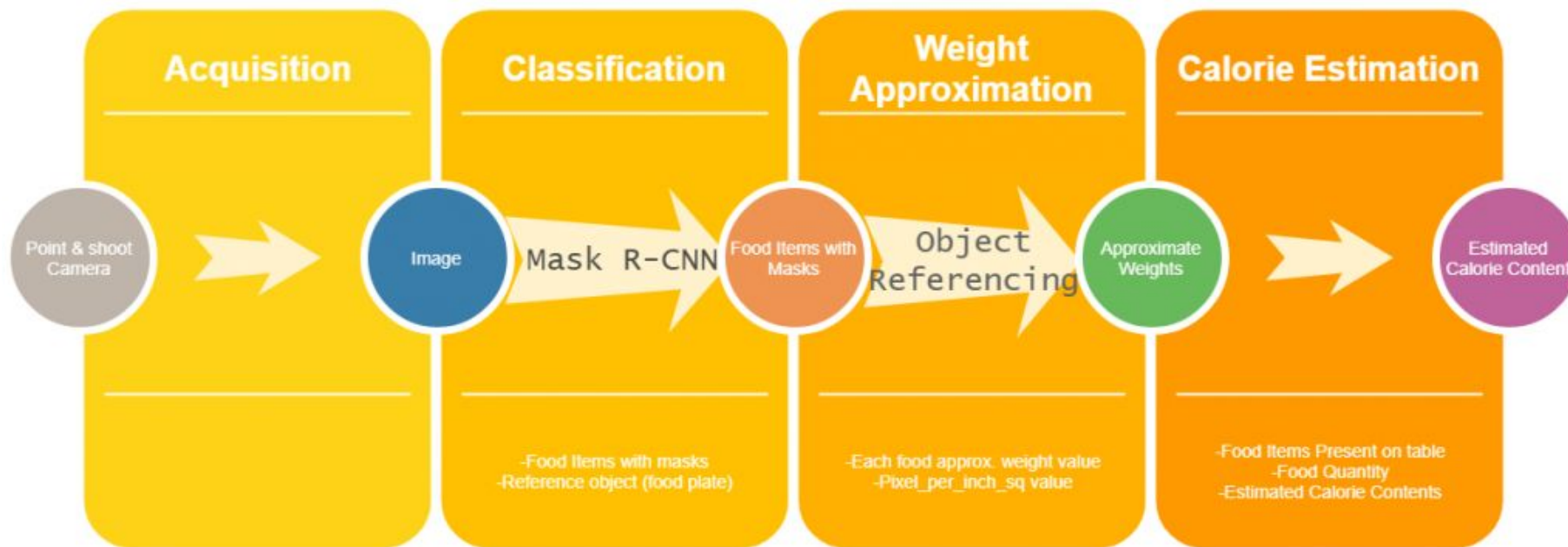
VS



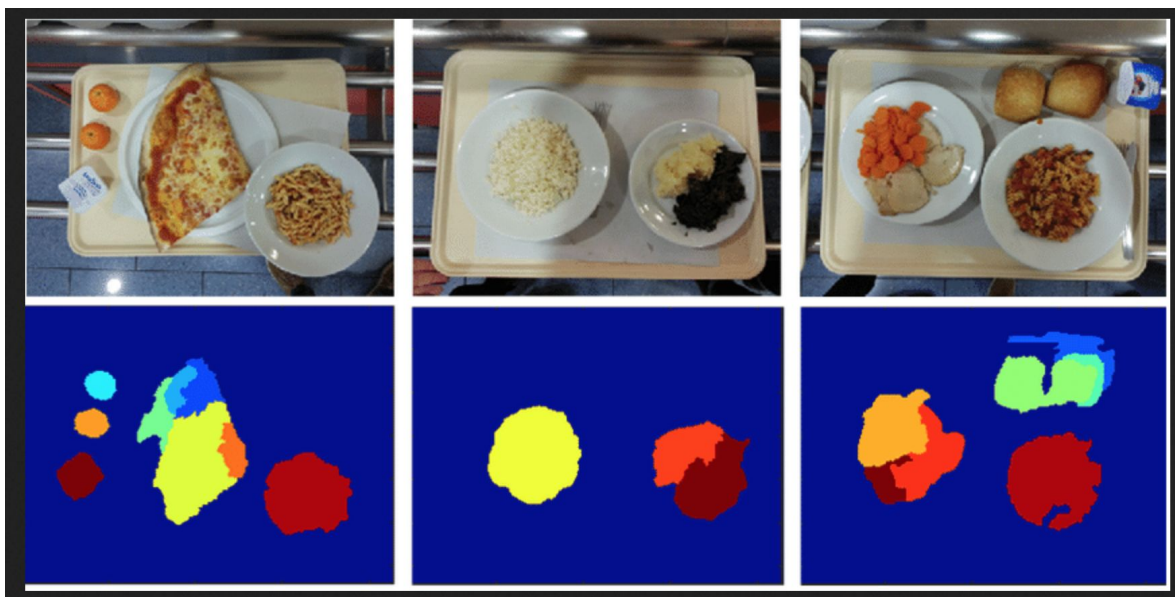
SOTA Food Recognition

Image-Based Calorie Estimation using Deep Learning

Kitamura, Aizawa et al 2009
Aguilar et al., 2017.



Methodology - dataset image crawling



Public dataset: Unimib2016

- Set of 1027 canteen food tray captured from top view
- 3616 food instances
- 73 food classes
- Polygonal boundaries
- 79% accuracy using CNN

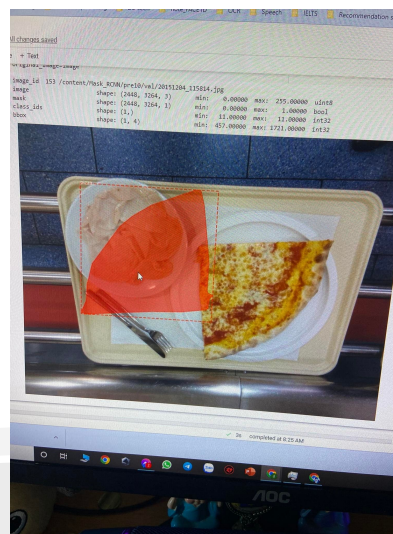
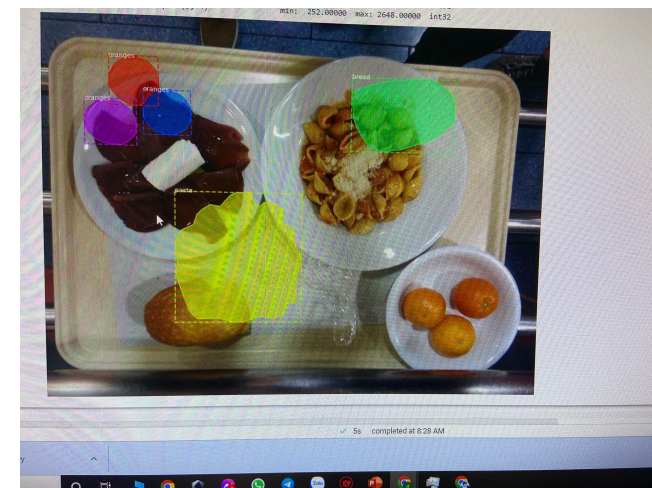
Data Preprocessing

image labelling

Issues regarding to data treatment.

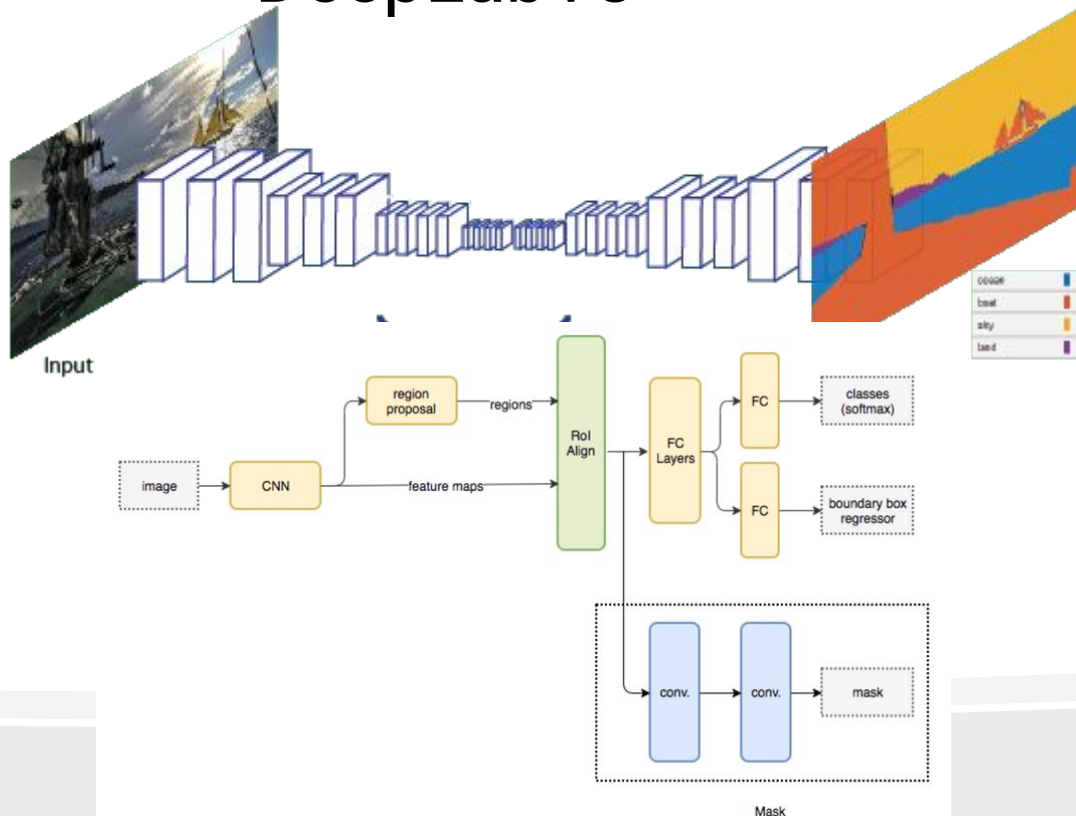
extension .mat -> .json

image rotation



Methodology - Model Training with Colab Pro

- Unet
- Mask RCNN
- DeepLabV3

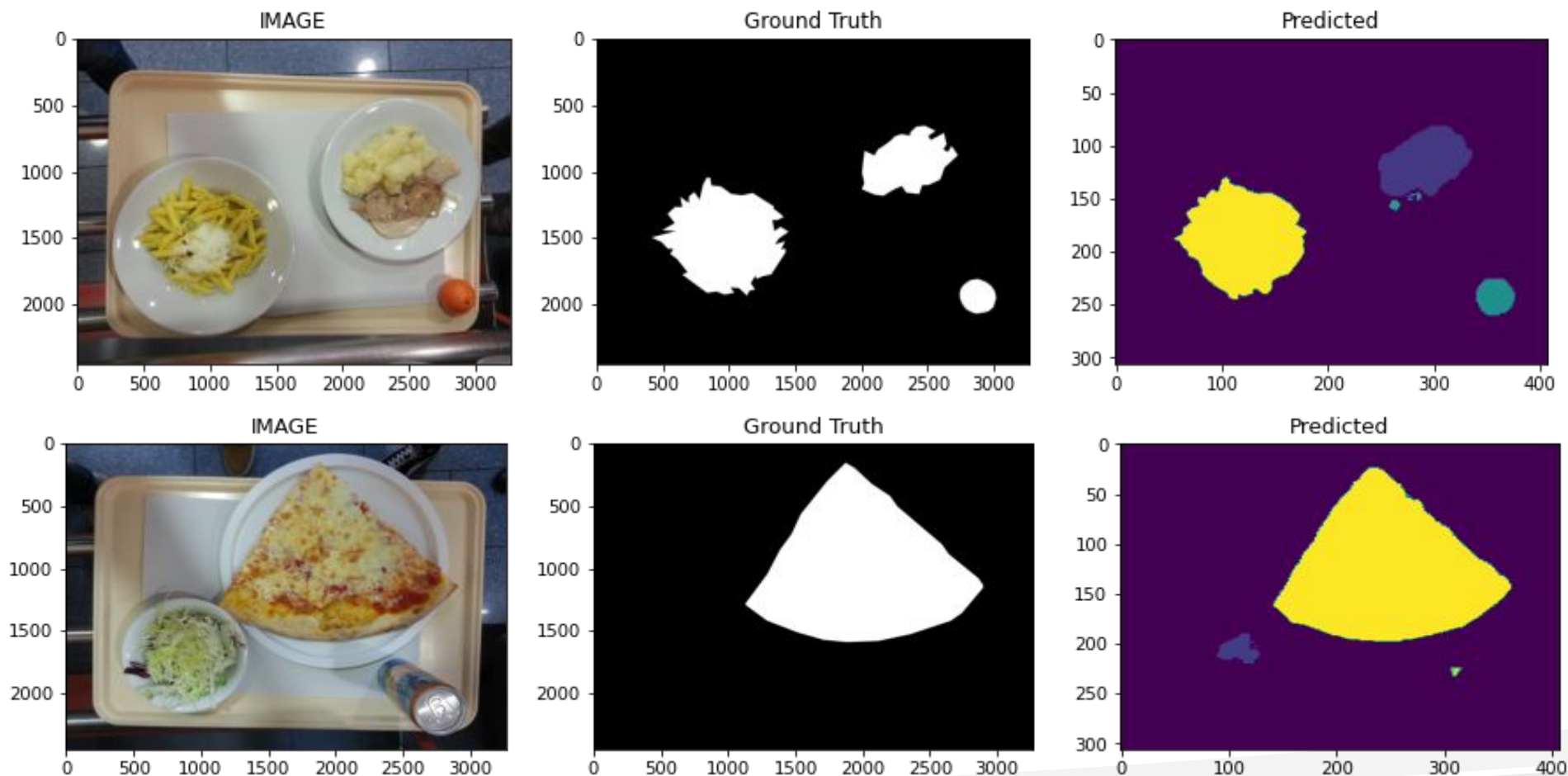


Training detail for
Mask RCNN
SGD ,
optimizer learning rate
batch size
epoch
pretrained models

Evaluation metric
COCO
mAP



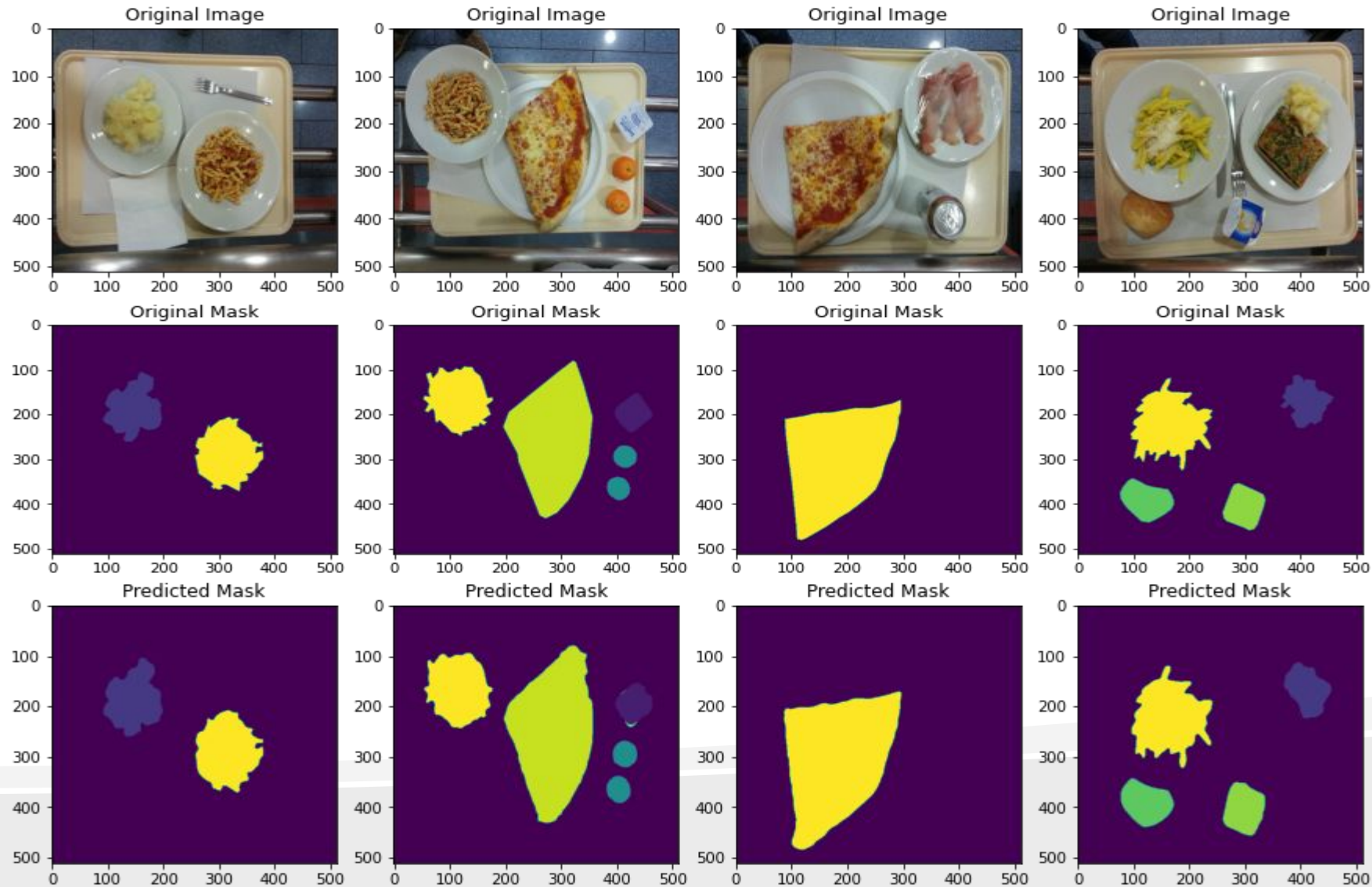
Results / Discussion (**Unet**from : Loss 0.05) 12hours



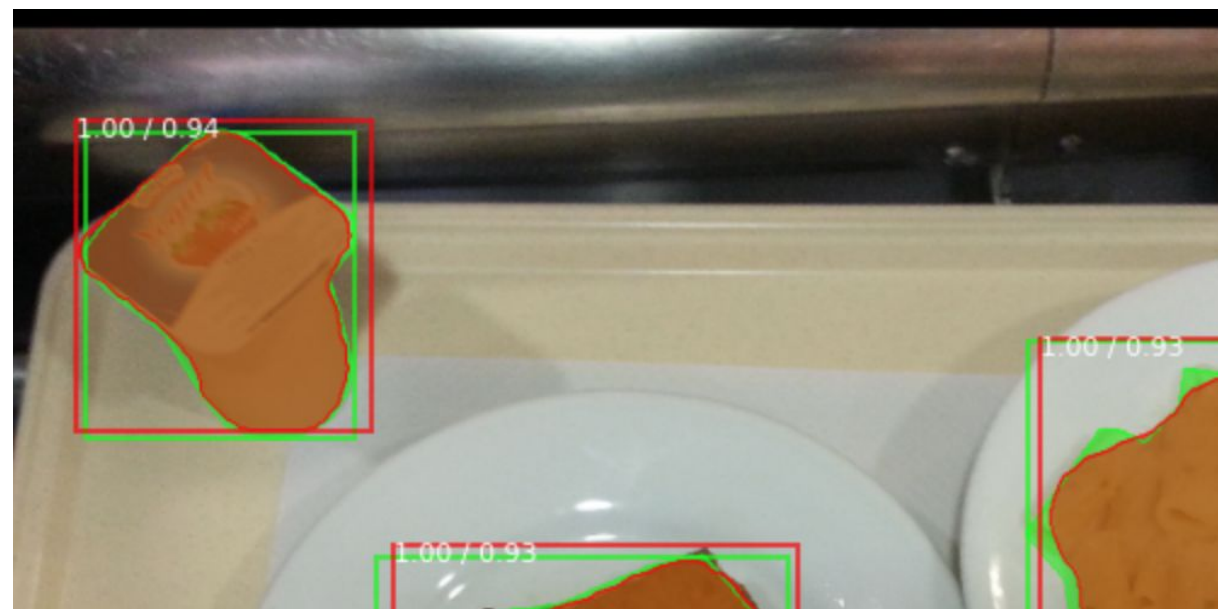
Loss Train: 0.02

Loss Val: 0.09

Results / Discussion (**DeepLab-V3+**)



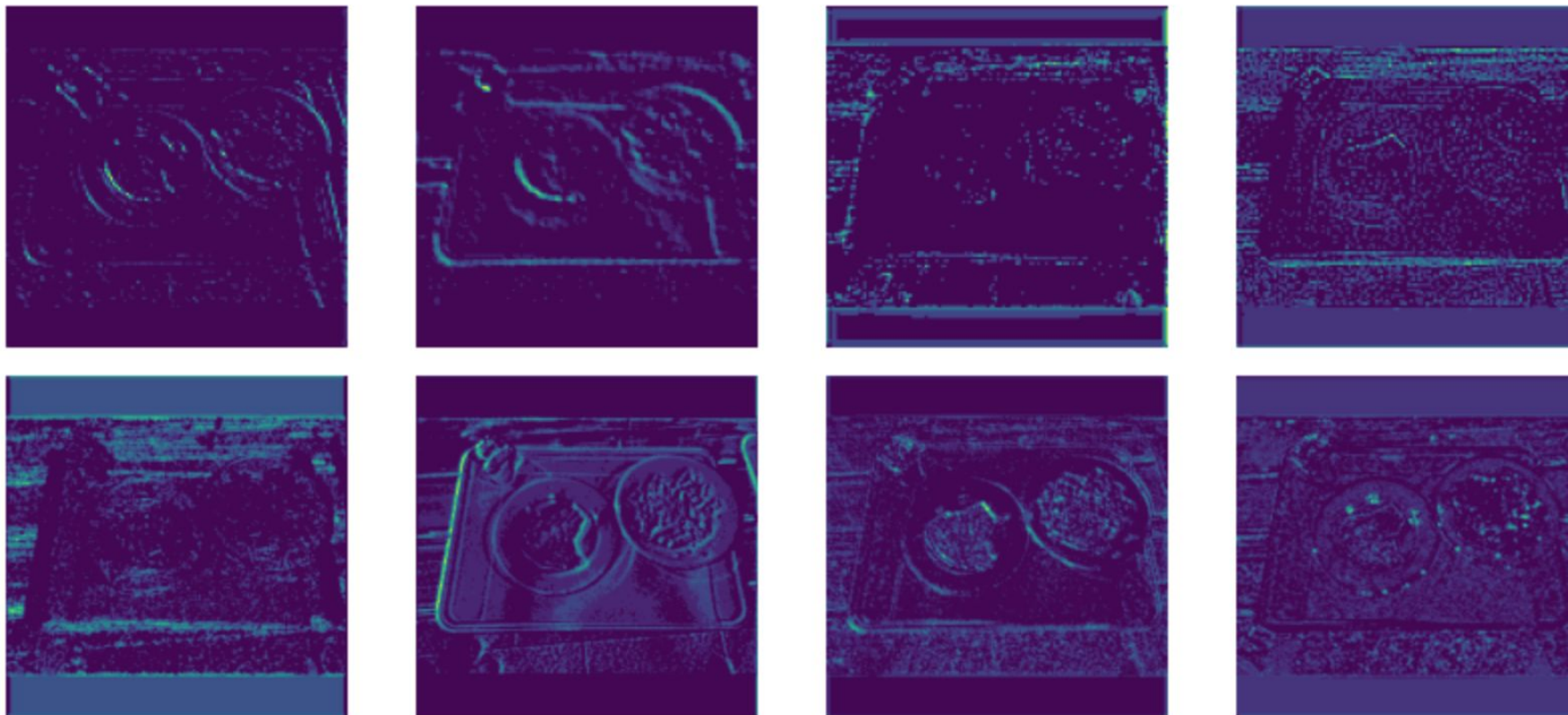
Results / Discussion (MaskRCNN: mAP: 0.903)



GT: Green | Pred:Orange | Caption =score/IOU

Mask RCNN Activation Layers

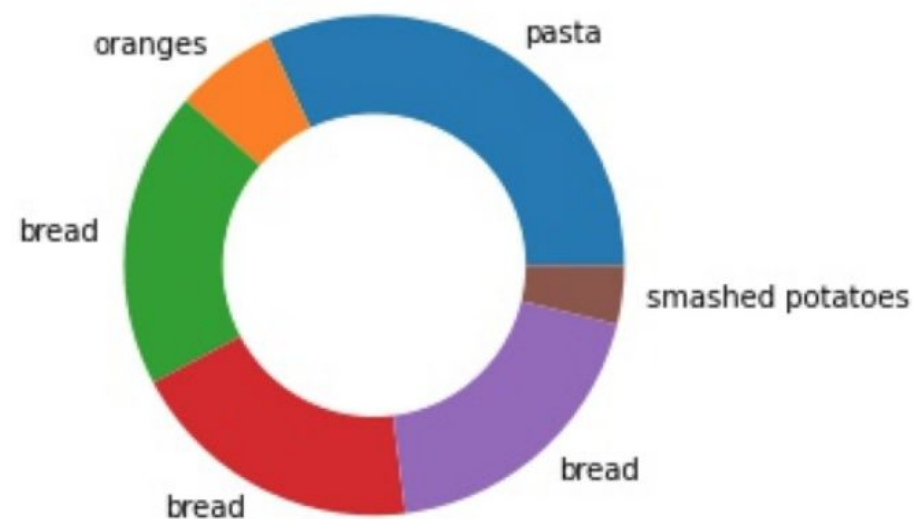
input_image	shape: (1, 1024, 1024, 3)	min: -123.70000	max: 151.10001	float32
res2c_out	shape: (1, 256, 256, 256)	min: 0.00000	max: 23.40121	float32
res3c_out	shape: (1, 128, 128, 512)	min: 0.00000	max: 27.75632	float32
rpn_bbox	shape: (1, 261888, 4)	min: -10.78870	max: 44.14271	float32
roi	shape: (1, 1000, 4)	min: 0.00000	max: 1.00000	float32



Application

$$PixelsPerInchSQ = \frac{MaskedPlatePixel}{RealPlateArea}$$

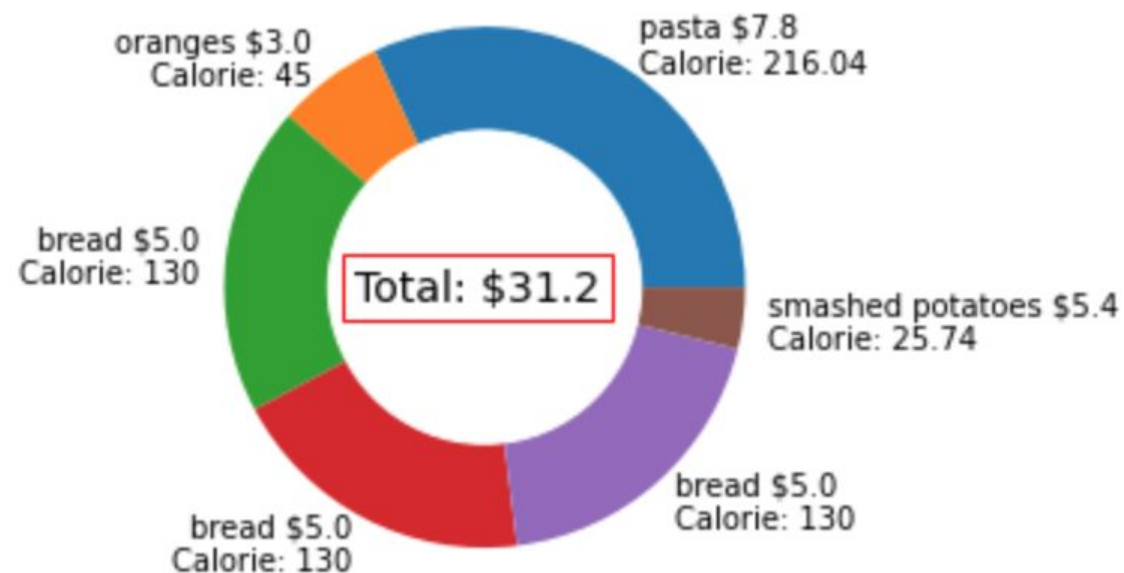
$$RealFoodArea = \frac{MaskedFoodPixel}{PixelsPerInchSQ}$$



Application

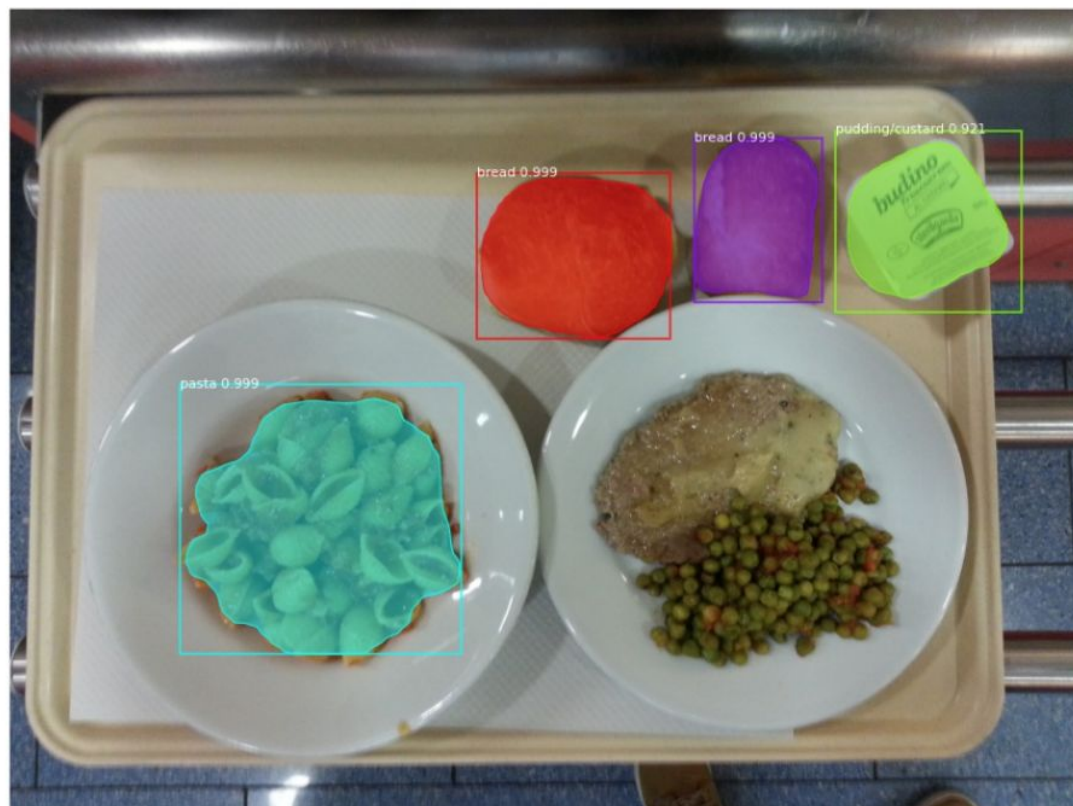
```
pasta with 216 calories  
oranges with 45 calories  
bread with 130 calories  
bread with 130 calories  
bread with 130 calories  
smashed potatoes with 25 calories
```

Dict: 13 labels | Price + Calories



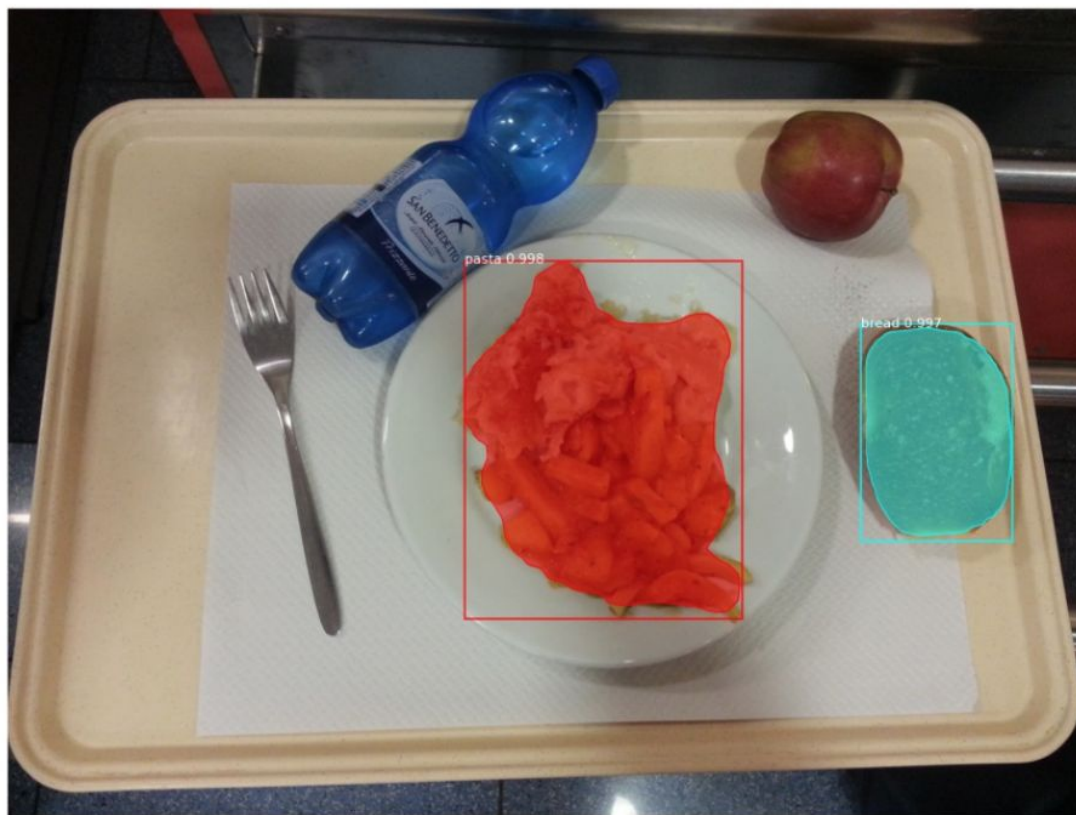
Application

bread with 130 calories
pasta with 169 calories
bread with 130 calories
pudding/custard with 130 calories



Application

pasta with 217 calories
bread with 130 calories



pasta \$7.8
Calorie: 217.82



bread \$5.0
Calorie: 130

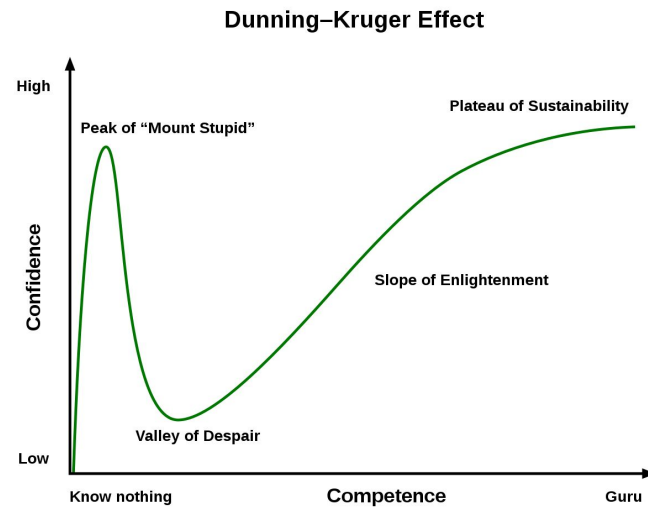
Application/ Not in dataset



pizza \$3.8
Calorie: 40.17



Conclusion



- Experience a DL project from scratch
- Define problem
- Source of Data
- Project Management







References

UNIMIB2016 Food Database :

<https://mldta.com/dataset/unimib2016-food-database/>

MarkRCNN: https://github.com/matterport/Mask_RCNN

<https://www.digitalocean.com/community/tutorials/how-to-make-a-web-application-using-flask-in-python-3>

<https://www.segmentationstudyguide.com/understanding-market-segmentation/market-segmentation-examples/market-segmentation-example-fast-food/>