

## Summary

YourFoodFriends is a segmentation and multi-class classification ensemble for detecting foods on a plate.

Masks of potential food candidates were generated via MetaAI's Segment-Anything-Model. With the masks, cropped images were inputted to a fine-tuned InceptionResNet V2 convolutional neural network for labeling.

From these outputs, the original food plate image is labeled with predicted food names.

## Problem Statement

Millions of Americans are visually impaired and might have trouble identifying food placed in front of them. Tourists in new countries might not know a dish and are too embarrassed to ask what it is.

With a simple picture, YourFoodFriends offers an aid to accurately identify the foods on a plate or tray. Trained on 101 foods ranging from all cultures and cuisines, our classification ensemble seeks to help the visually impaired and tourists in need.

## Data Used For Classification

### Food-101 (Bossard et al.)

- Contains 101 different labels with 1000 images each
- Pre-split into 75,750 training and 25,250 validation samples
- Training samples contain deliberate mis-labelings



Dumplings

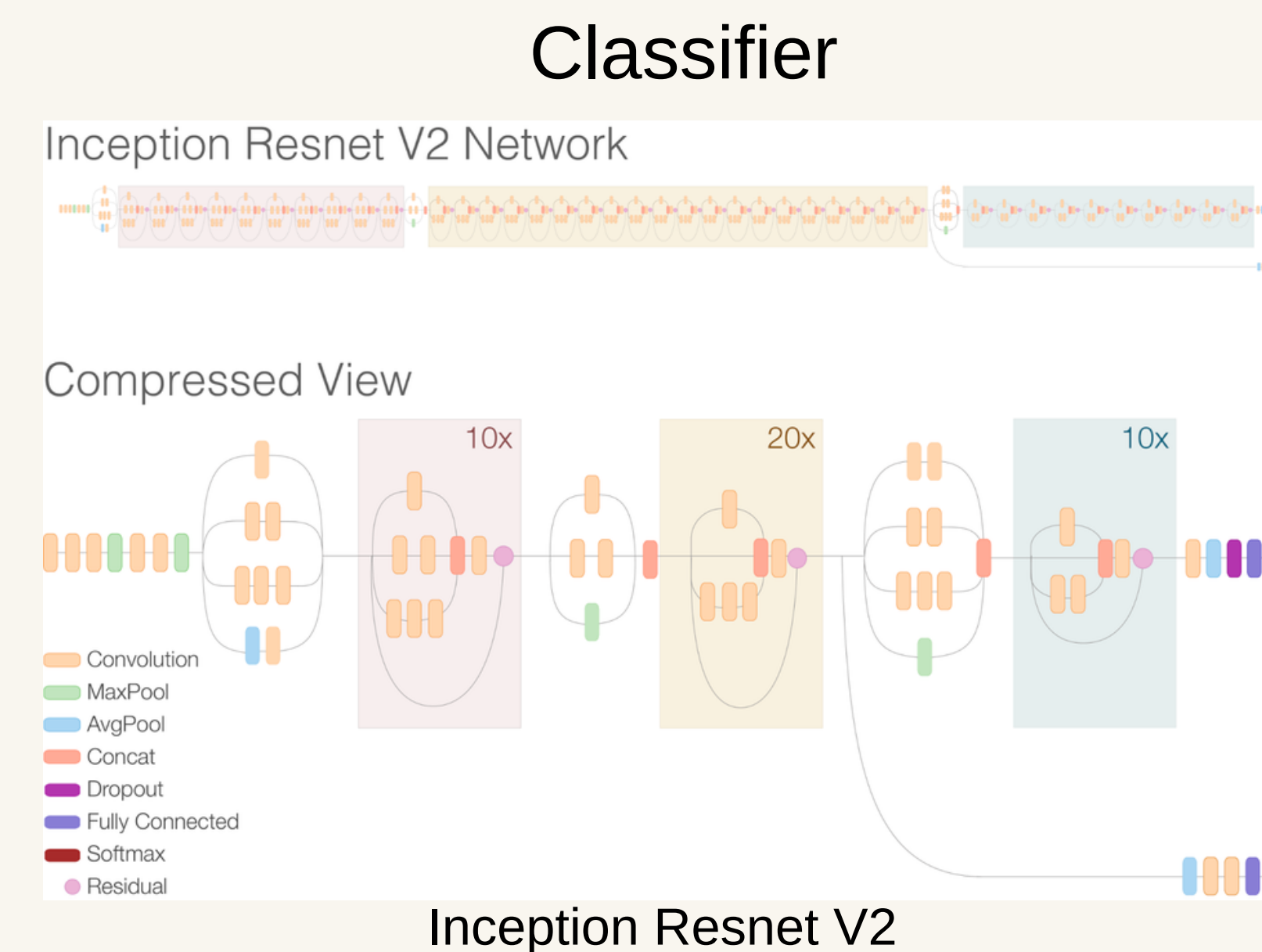
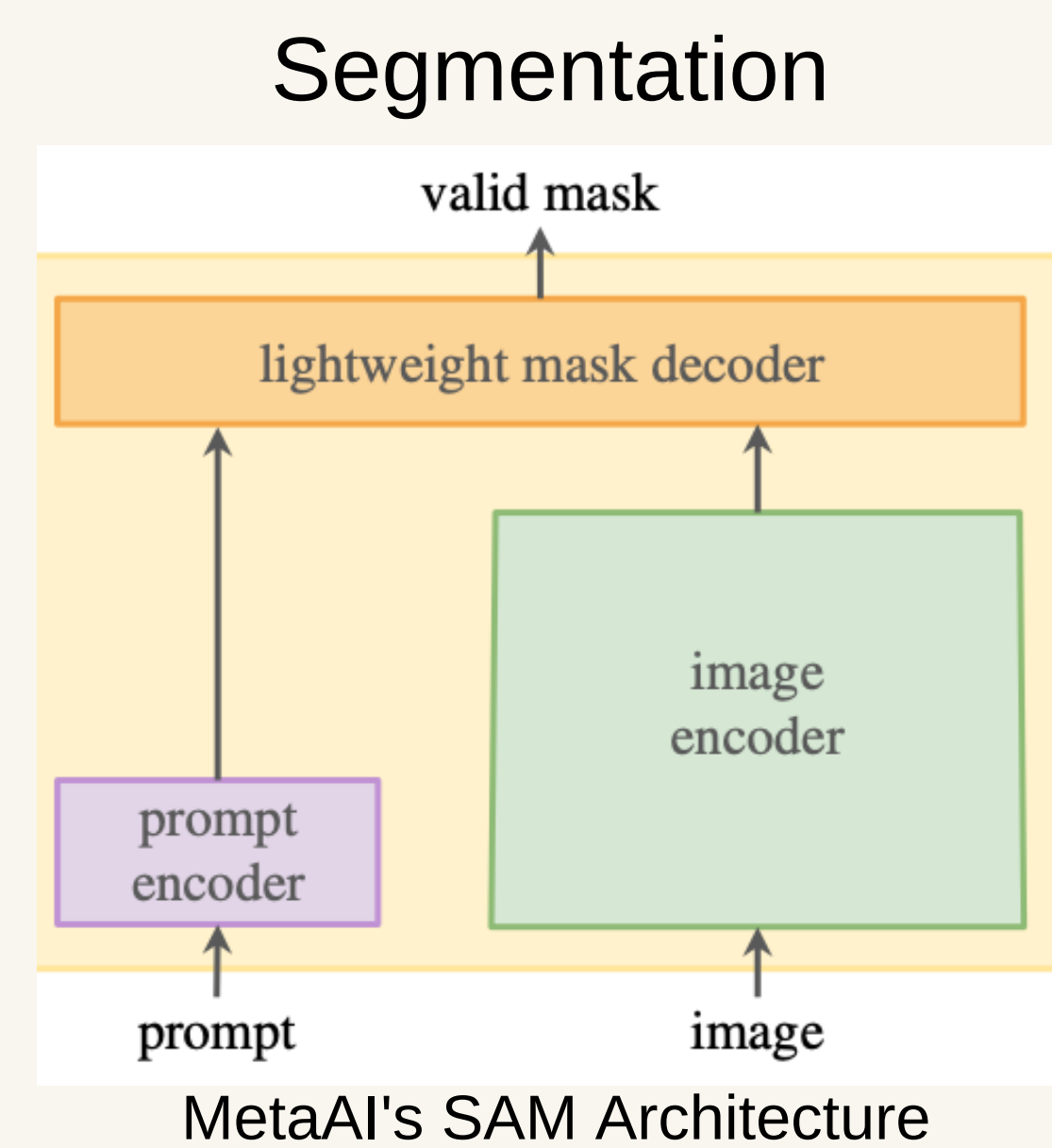


Pancakes



Ice Cream

## Algorithm Specifications



## Architecture

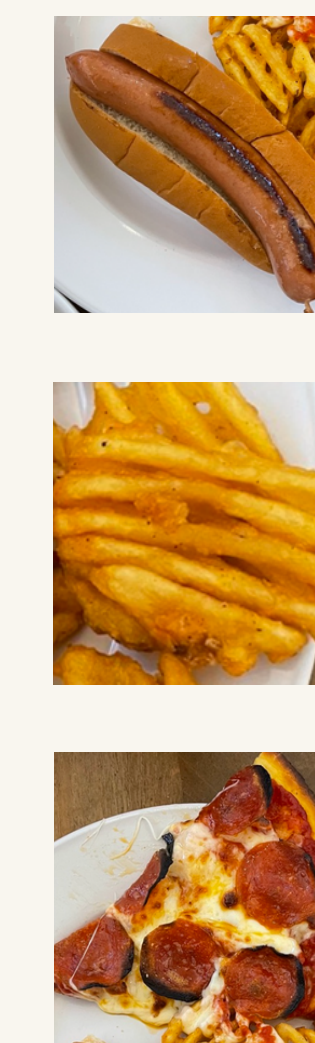
Input RGB image of food tray



**SAM**



Segmented images



**Our Classifier**



Output food labels

Hotdog

French Fries

Pizza

## Results

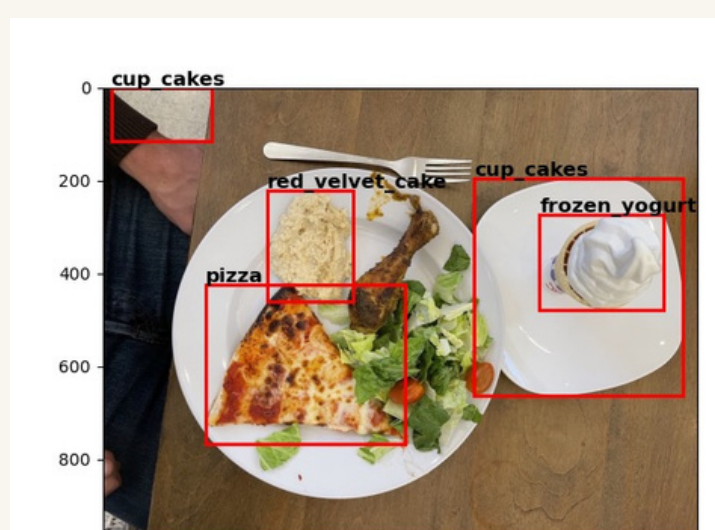
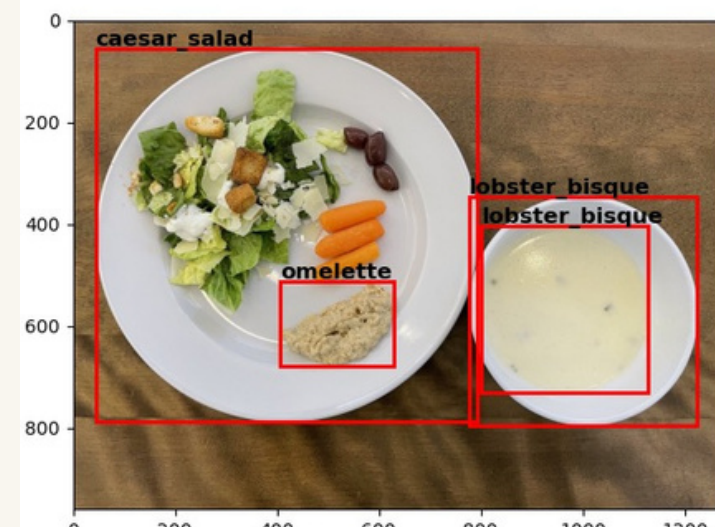
Original Ratty plate of food



SAM mask unfiltered output



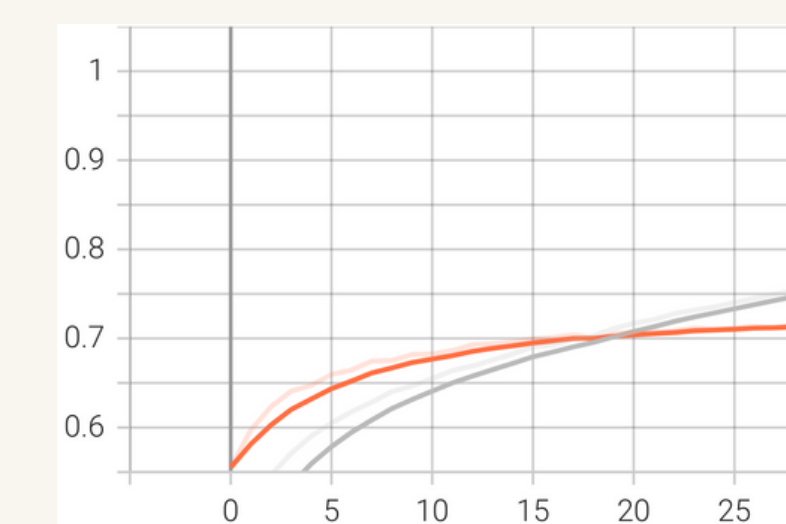
Output classified food tray



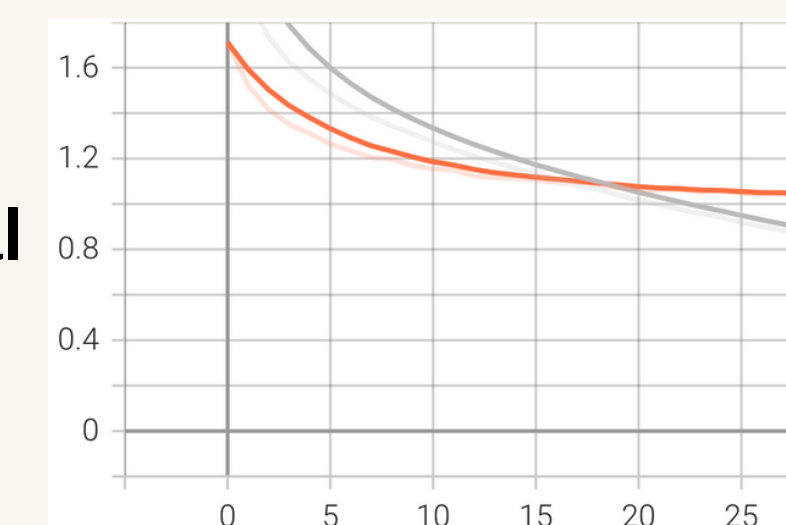
## Classifier Accuracy

Model #5 with IRNv2 Base

Accuracy for Train and Test per Epoch

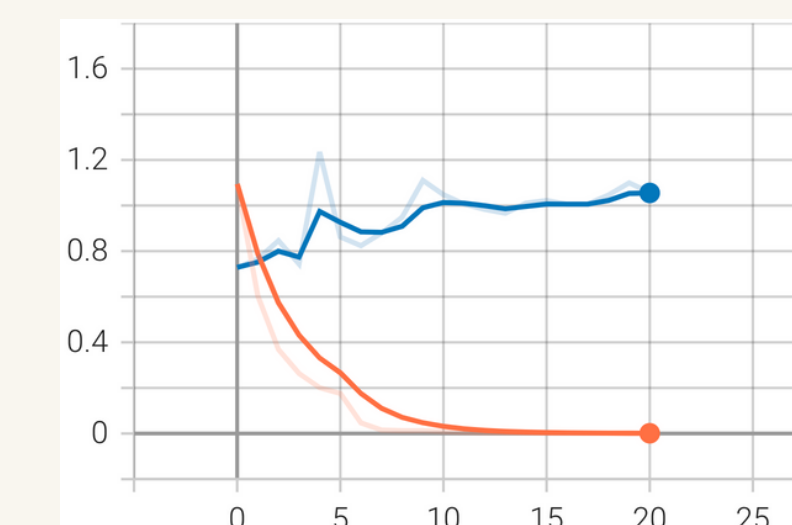
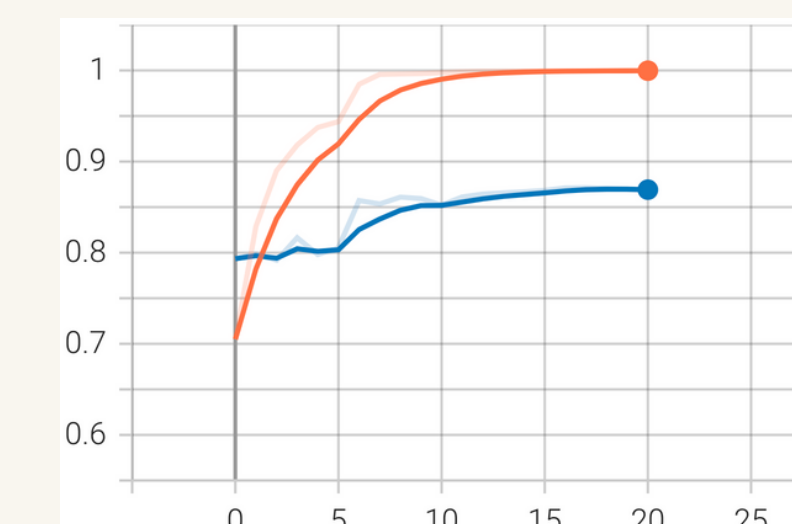


Sparse Categorical Loss per Epoch



Gray: Training Accuracy  
Orange: Validation Accuracy

Model #5 Fine Tuning



Orange: Training Accuracy  
Blue: Validation Accuracy

Various Classification Model Performances

Model #	Base	Head	Train Acc	Test Acc
1	VGG16	f,4096,d05,d096,d05	0.4645	0.5023
2	Xcept	g	0.7007	0.6727
3	FT MNv2	g	0.9999	0.7895
4	FT ENB3	g	0.9998	0.8265
5	FT IRNv2	g,2048,d05	0.9997	0.8713

Key: f=flatten; g=GlobalAveragePooling2D(); # = Dense(#)  
d05 = dropout(0.5); FT = fine-tuning

## References

- [1] Martin Abadi et al. TensorFlow: Large-scale machine learn-ing on heterogeneous systems, 2015. Software available from tensorflow.org
- [2] Lukas Bossard, Matthieu Guillaumin, and Luc Van Gool. Food-101 – mining discriminative components with random forests. In European Conference on Computer Vision, 2014
- [3] Alexander Kirillov et al. Segment anything. arXiv:2304.02643, 2023
- [4] Karen Simonyan and Andrew Zisserman. Very deep convolu-tional networks for large-scale image recognition, 2015
- [5] Christian Szegedy, Sergey Ioffe, and Vincent Vanhoucke. Inception-v4, inception-resnet and the impact of residual con-nections on learning. CoRR, abs/1602.07261, 201

## Acknowledgements

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