

AISS CV – Group 6 – Food Label Recognition

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- √ Wo kommt es her?
- Wo wurde es verarbeitet?
- ✓ Wie hoch ist der regionale Anteil?

Neutral geprüft durch: Kontroll GmbH

ÖKO*TEST

ÖKO-TEST-Magazin



Bioland





















nach EG-Öko-Verordnung

QMilch QMilch.info

























VEGAN

Our Solution



Our CV application detects labels on supermarket items to:

- Help users differentiate between legitimate & marketing labels
- Provide additional information on legitimate labels







Quickly assess properties and quality of different products while directly engaging with them in the supermarket

Developing Our Solution



- Collecting & Preparing Data
- Choosing & Training The Model
- Evaluating our Model
- Putting It All Together

Creating Our Data Set





No suitable data sets publicly available for our use case



EDEKA

With label: 160 No label: 160



With label: 37 No label: 37

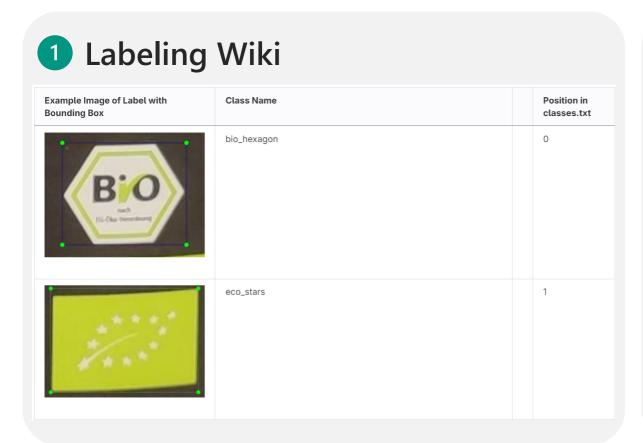


Label Photos

Tool: LabelImg

Using Labeling Guidelines To Ensure Consistency



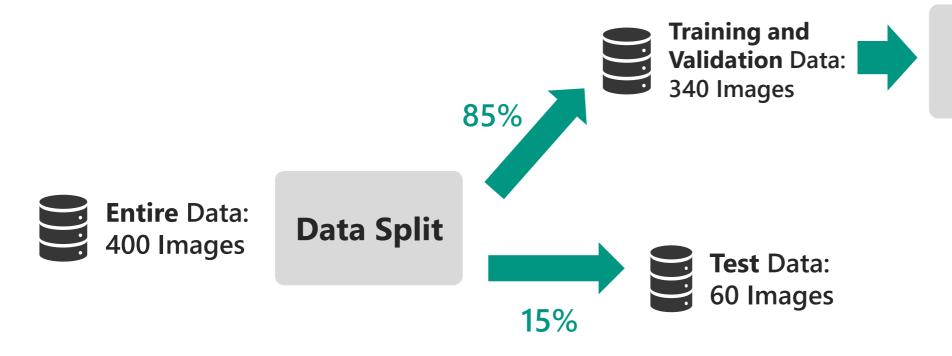




+ Best Practice Labeling Rules

Data Split and Augmentation





Data Augmentation

~1400 Images

Data Augmentation Operations



Original



Safe Rotate & Color Jitter



Contrast & Cropping



Horizontal
Flip
&
Blurring

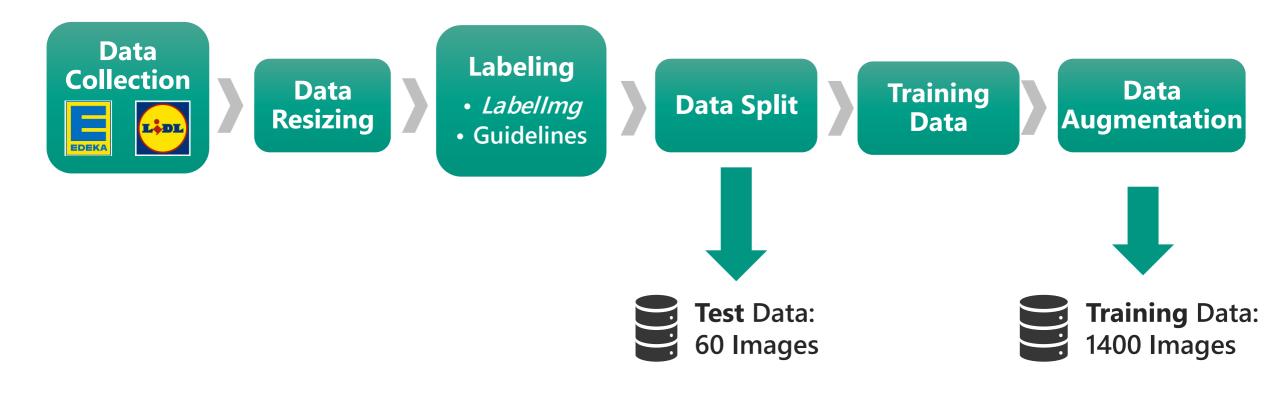


Vertical
Flip
&
Elastic
Transform



Our Data Processing Pipeline





Choosing The Right Model For The Task





Pretrained & SOTA object detection model



Small model for fast inference on Jetson Nano (Accuracy-Speed Tradeoff)



Simple training and inference API using "Darknet" Framework



Tiny YOLO v4 Top + Darknet





Final Training & Using Darknet



Train on separate GPU for acceleration

1771 train images
NVIDIA RTX 4090 GPU



Deploy and Use on Jetson Nano





To run Darknet, it's necessary to recompile OpenCV in order to support CUDA and GStreamer

Good Quality of Model Predictions











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Correct: 188



Missed: 26

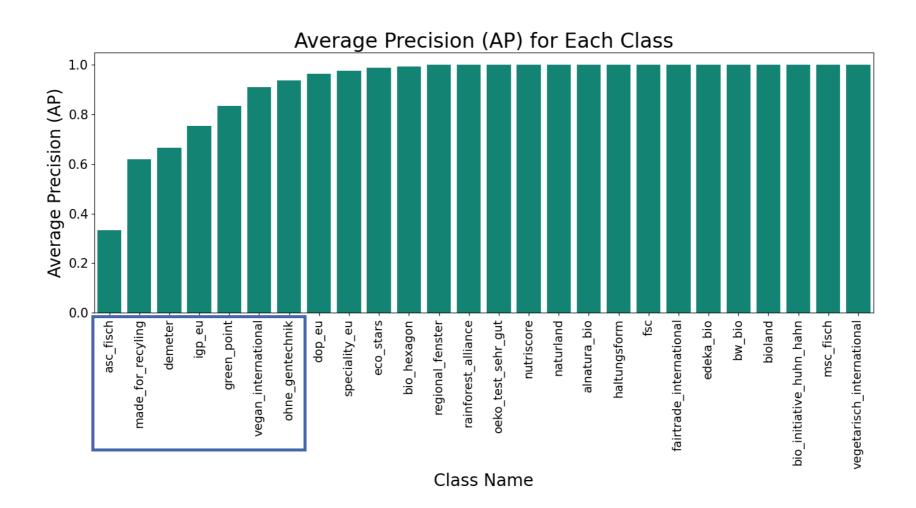


no label Incorrect: 15

Our Model Achieves A High Average Precision

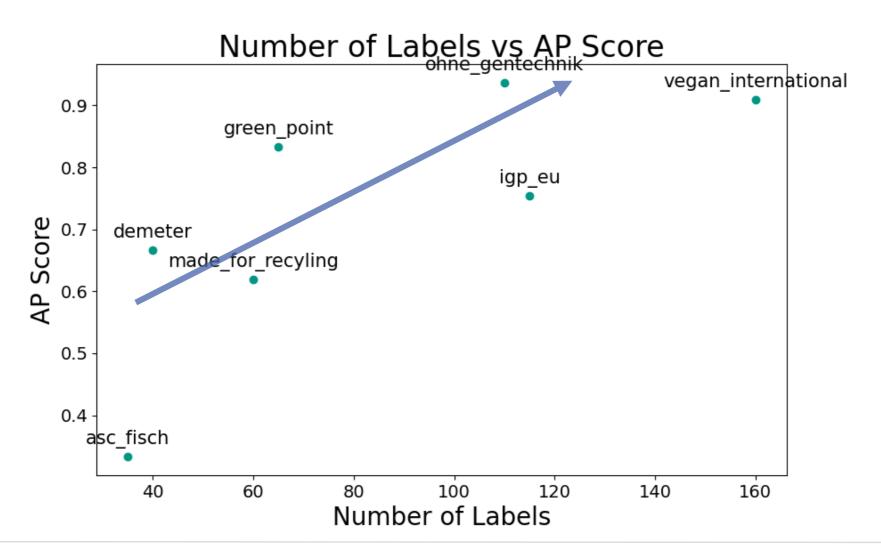
13





Relationship: Number of Labels & AP Score

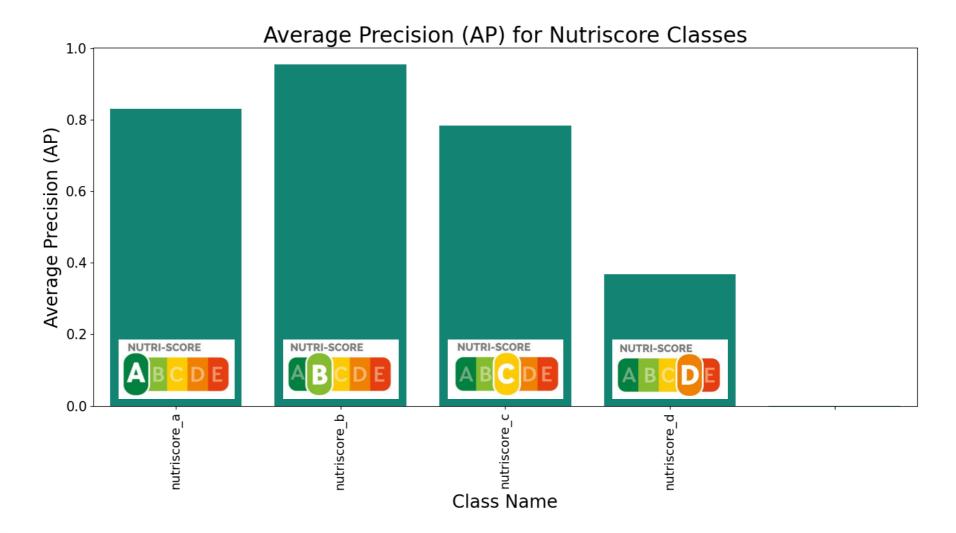




Improving Poor Performing Classes

15

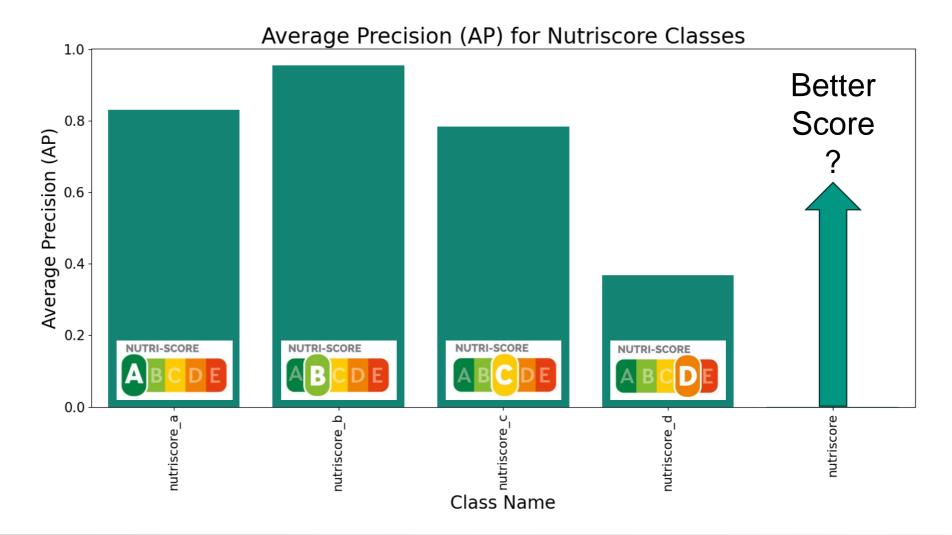




Improving Poor Performing Classes

16

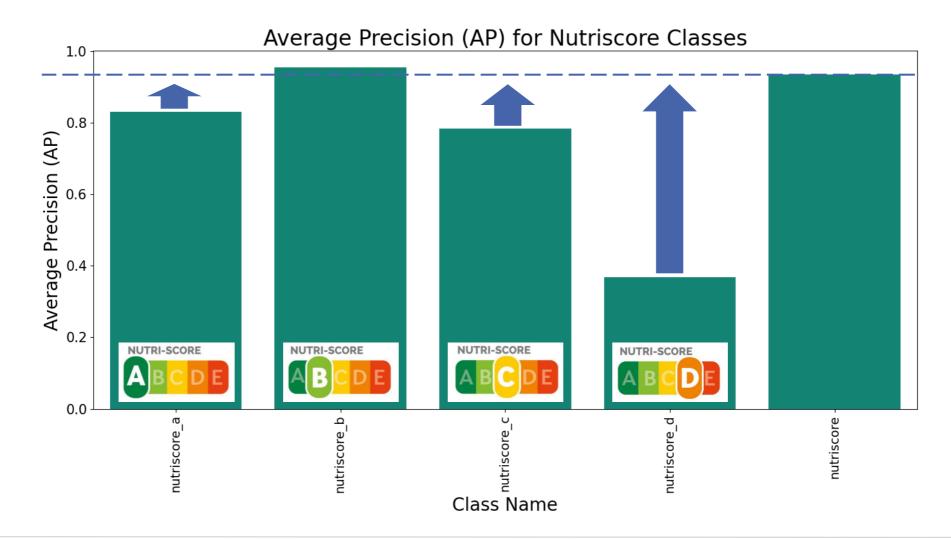




Improving Poor Performing Classes

17

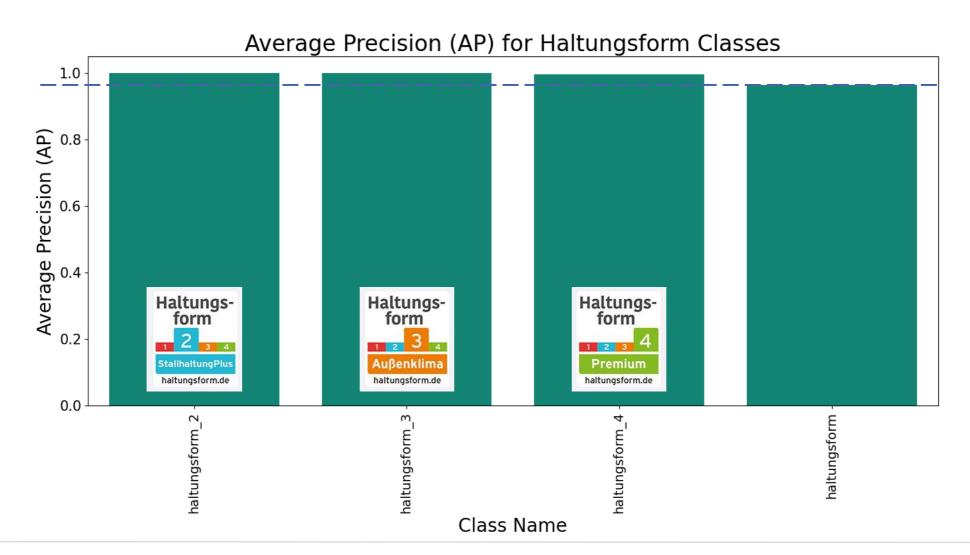




No Benefit For Already Well Performing Classes

18

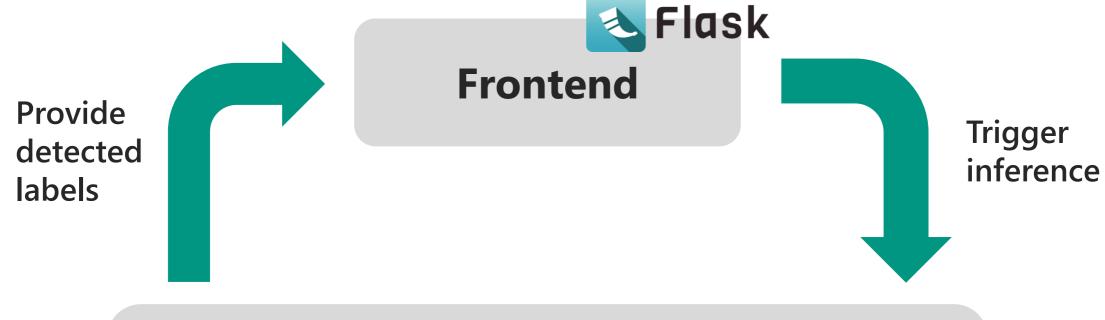




Putting It All Together

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Let's Run Our Model!

Thank You For Your Attention ©





Business Case

 CV Application to detect labels on supermarket items and provide information to users



Data Preparation

- Collect images from supermarket
- Augment images to increase model performance → avoid data leakage



Training & Inference

 Train Darknet on GPU, run inference on Jetson Nano



Frontend Integration

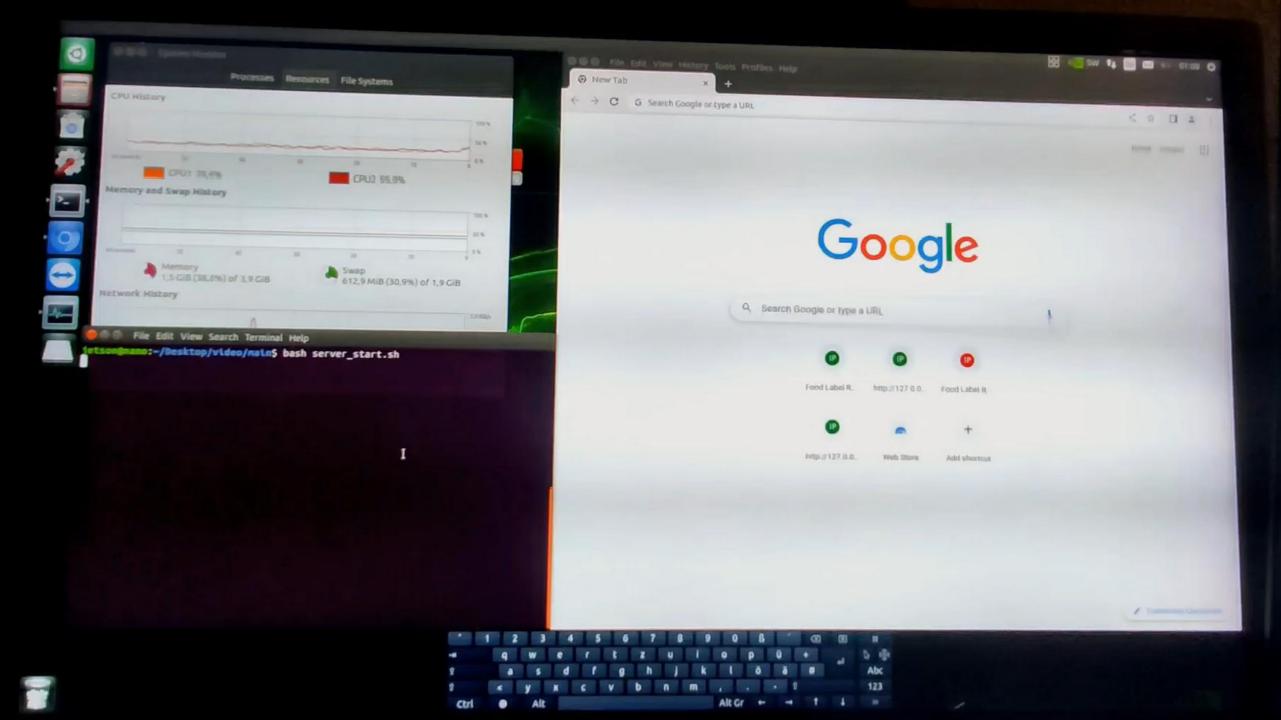
Flask webapp allows users to run inference on Jetson Nano using video/ image stream from CSI camera



Backup Slides

26 July 2024 IISM / KSRI@KIT

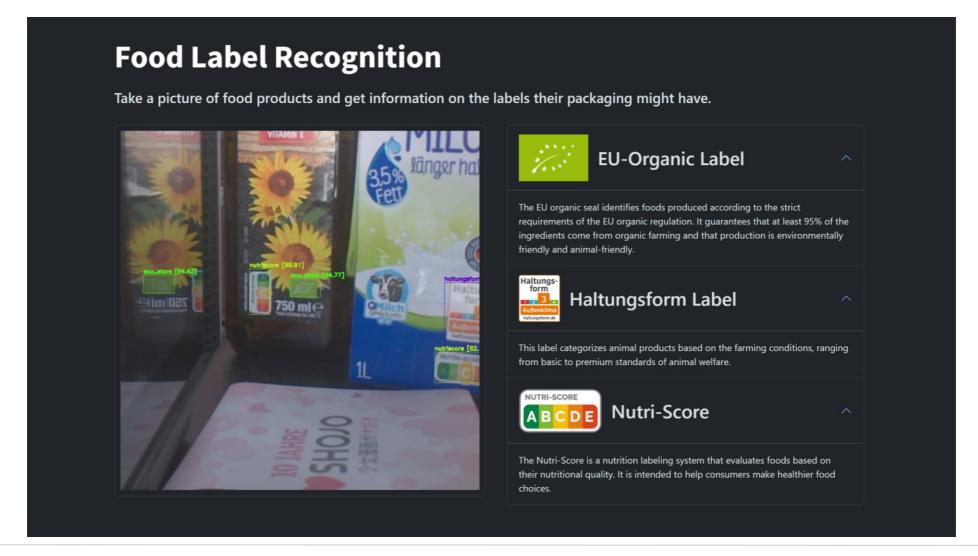
22





24





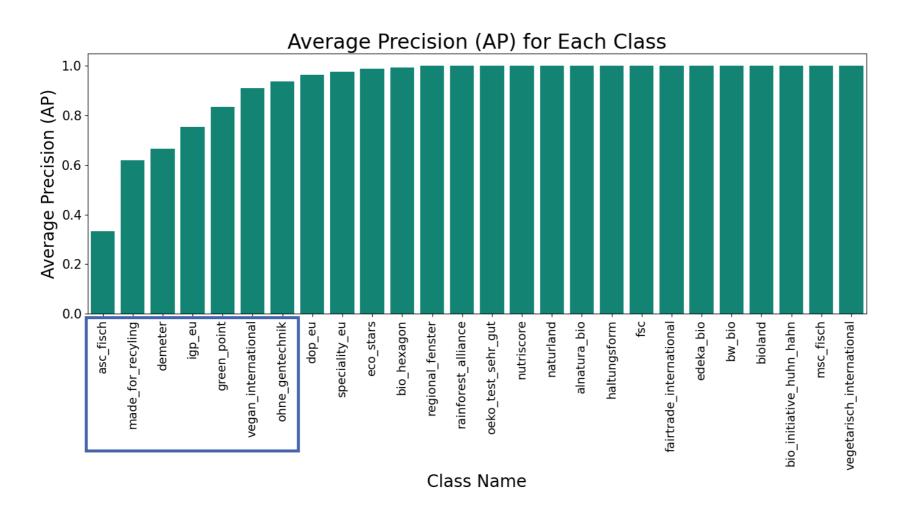
Results in Detail



- Mean Average Precision Score: 92%
- Intersection over Union: 74.06 %
- Test BBoxes: 214
 - **TP: 188**
 - FP: 26
 - FN: 15

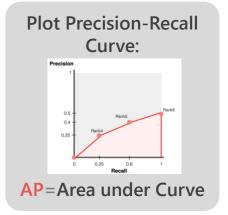
Our Model Achieves A High Average Precision





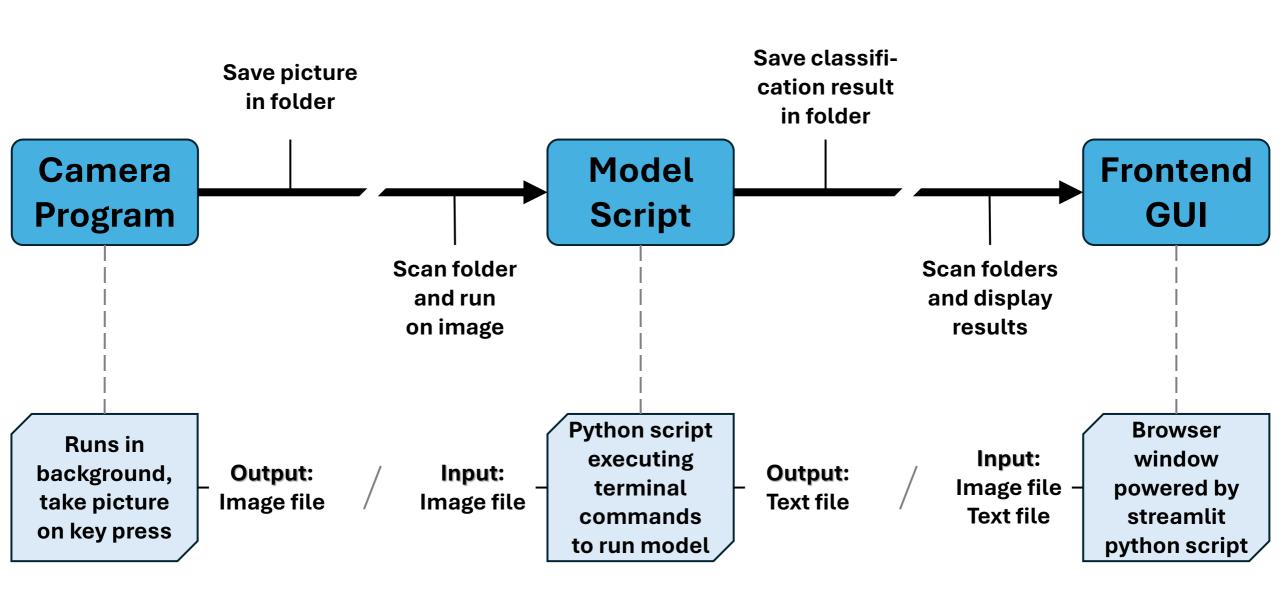
AP per class:

Calculate Precision and Recall based on IoU for different confidence thresholds

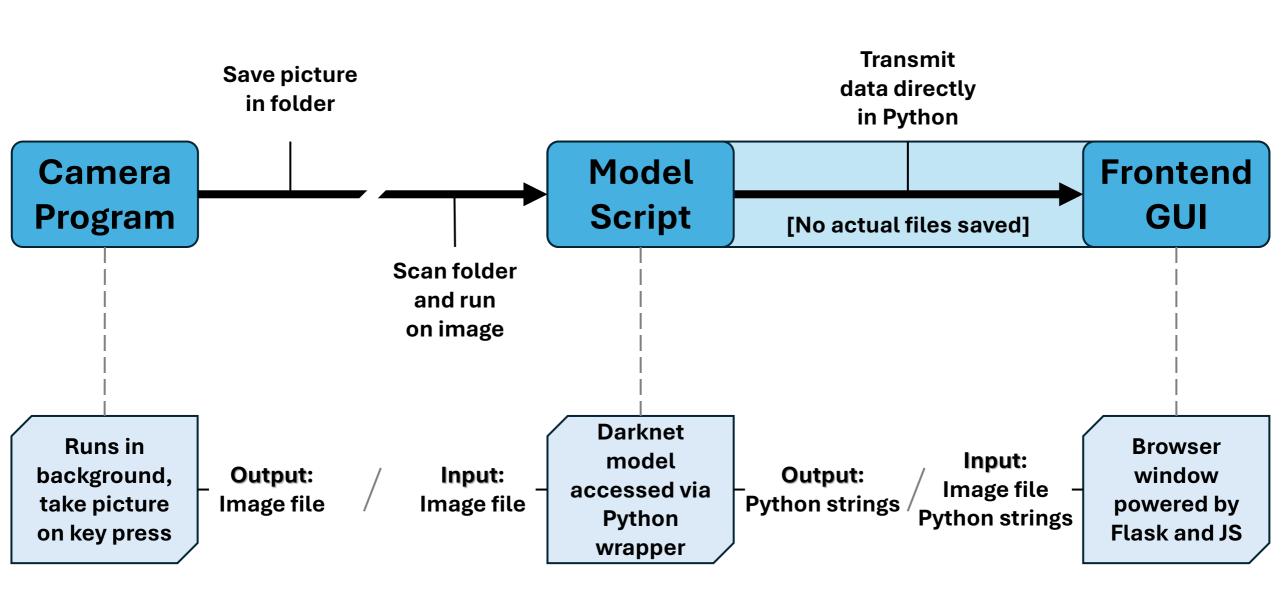


AP considers the precision-recall trade-off

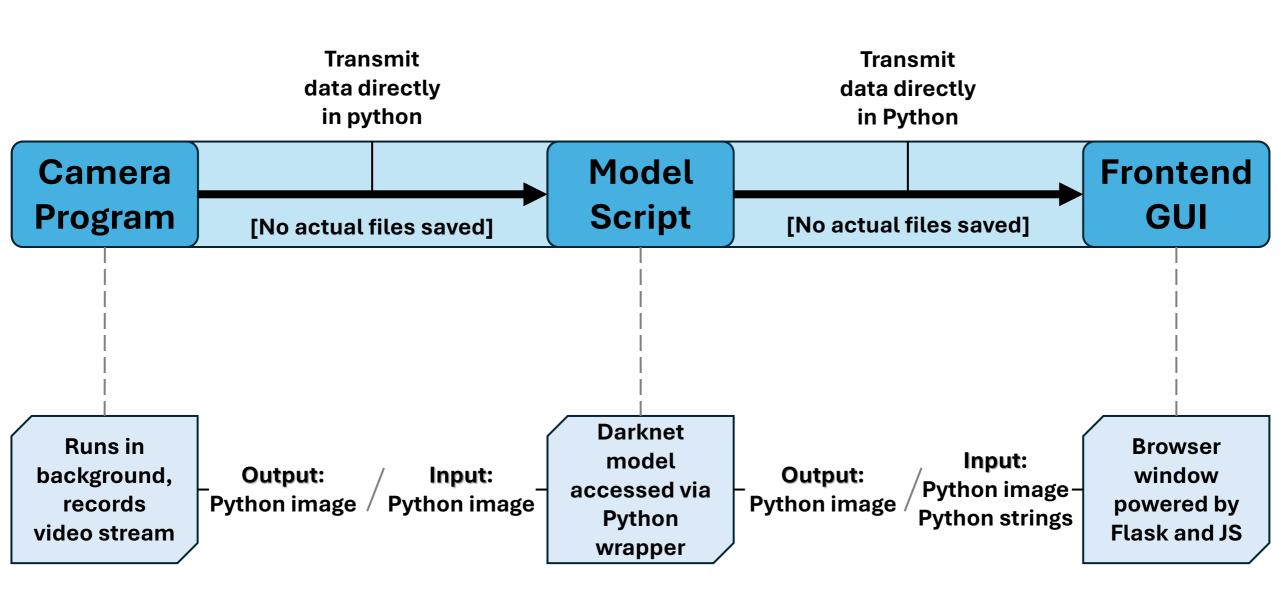
Program Flow – 1st Version with Streamlit



Program Flow – 2nd Version: Flask Images



Program Flow – 3rd Version: Flask Video



References



- https://www.freepik.com/free-vector/hand-drawn-woman-supermarket_4104602.htm#fromView=search&page=2&position=0&uuid=215175ba-a29a-4a77-aeb5-fb61ce7917c2
- https://www.evidentlyai.com/ranking-metrics/mean-average-precision-map
- https://www.verbraucherzentrale.de/wissen/lebensmittel/kennzeichnung-und-inhaltsstoffe/nutriscore-was-bedeutet-die-kennzeichnung-76209