

SYNOPSYS[®]

2022 Synopsys ARC AloT Design Contest - 決賽作品

Foodie Fortress

智慧冰箱管理系統

隊名: 709 板板之王

成員:朱育欣、謝瑞峰、林育萱

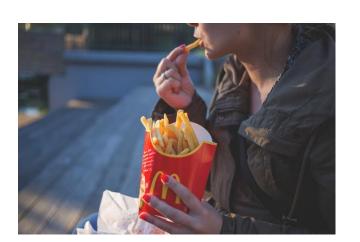
指導教授:張俊盛



User Pain Point

你的共用冰箱是不是有這種問題?

不小心誤吃到 別人的食物



被遺忘的食品 佔據空間、腐敗



出門購物時 忘記要添購的食材





FoodieFortress 智慧冰箱管理系統

Al+loT技術保護食品歸屬,記錄存取紀錄,提醒食品新鮮度解決共用冰箱誤取、過期食品和食材選購的問題。



Agenda

- 1. System Introduction
- 2. Model Overview
- 3. System Architecture
- 4. Conclusion





System Features

Food Classification

Use **CV model** to classify the food and keep record in DB.

Dash Board Display

Intergrate with DB and use **Line Bot** as **User Interface**, help user keep track on their own food.

Expire Food Notification

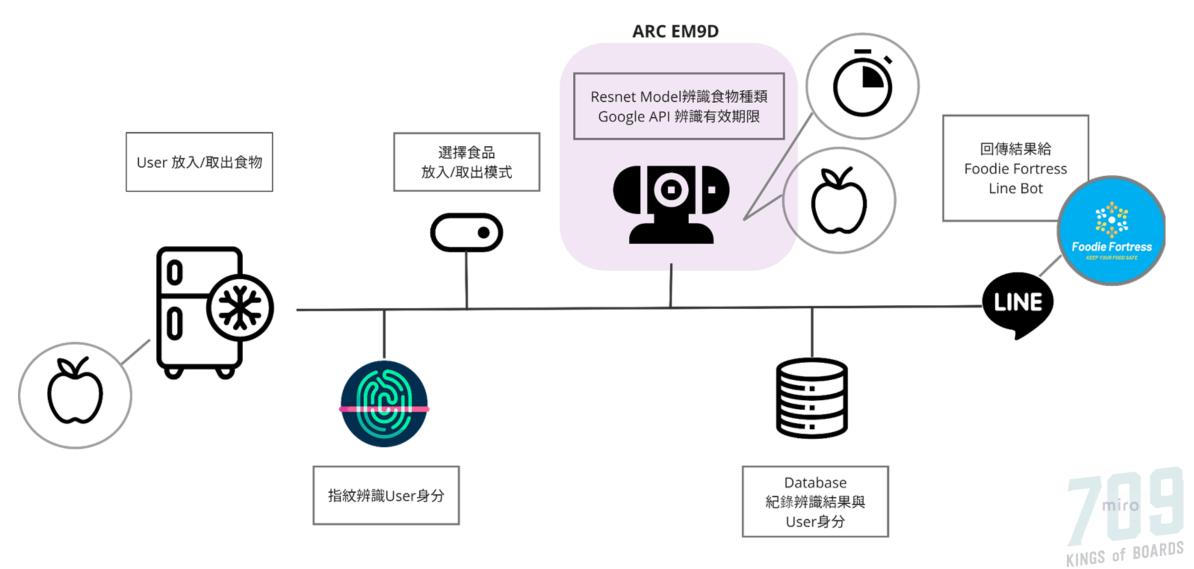
Use **Google Vision**(OCR) API to
detect expire date,
notify user when the
food goes bad.

User Authentication

Use **Fingerprint** and Line bot **token** to authenticate a user.



System Process



System Strengths

Efficiency

Multi-classification model which fits EM9d processor architecture well

Intergration

Intergrate with external resource, eg, **GCP**, **Line bot**

Security

Use **Fingerprint** to protect user identity and keep their food safe

Innovation

Use CV model to kill the pain point for shared fridge





Dataset: Fruits360 + Self-generated dataset

1. 以資料集為<u>Fruits360</u>為基礎

Kaggle public dataset, over 2600 upvote fruit pictures

- Take 10 class as below, and ~4.10 platures for some alone :

Apple

Banana

Corn

Eggplant

Guava

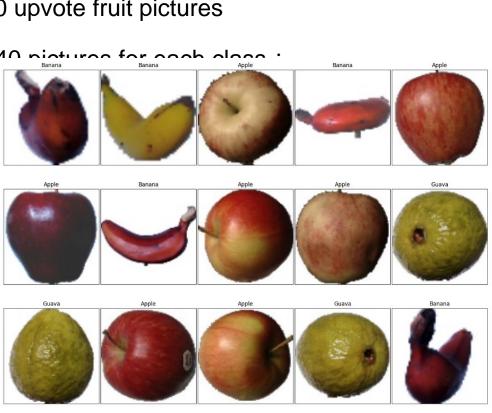
Lemon

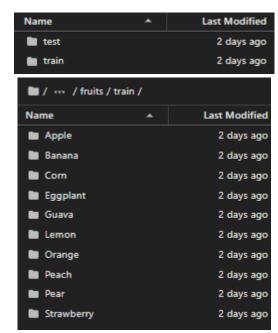
Orange

Peach

Pear

Strawberry



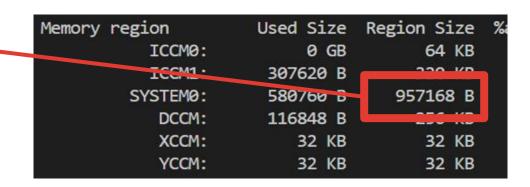




Difficulties

1. Memory Limitation: System0 < 957168 B

困境:大部分CV model 參數太多



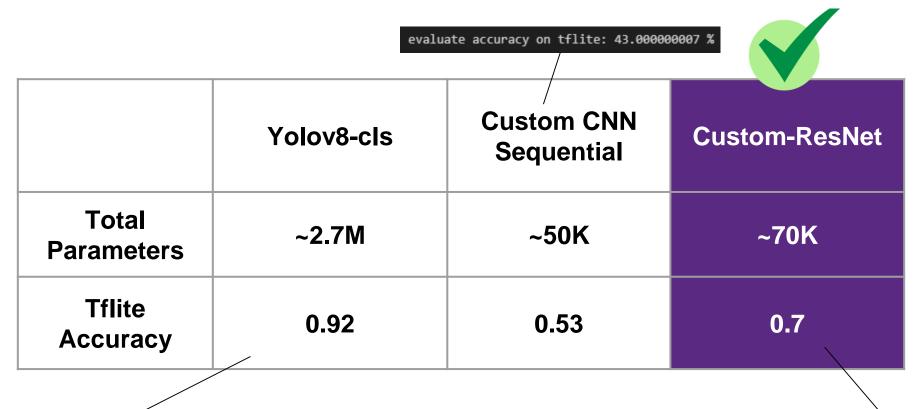
1. Low muli-class accuracy:

困境:黑白圖片難以達成食物辨識的多種Class分類

→ Custom ResNet Model



Classic CV Model Comparison



evaluate accuracy on tflite: 73.58%



Preliminary: ResNet

→ 解決Degradation Problem:

Convolution層 中間多開了直通連路(Skip Connection),將每一層的輸出改成 原本學習到的特徵 + 上一層的結果,解決 Degradation Problem。

→ 降低學習難度:

直通連路與卷積層 直接相加而非用連接的方式,將學習目標改為降低殘差(Residual),降低模型學習難度。

→ 減少模型參數量:

殘差塊使用 bottleneck形式,將3x3卷積核拆成1x1,3x3,1x1卷積核,可以在減少模型參數量的同時兼顧到更高的維度。

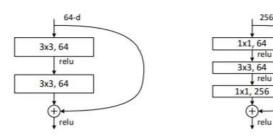


Figure 5. A deeper residual function \mathcal{F} for ImageNet. Left: a building block (on 56×56 feature maps) as in Fig. 3 for ResNet-34. Right: a "bottleneck" building block for ResNet-50/101/152.

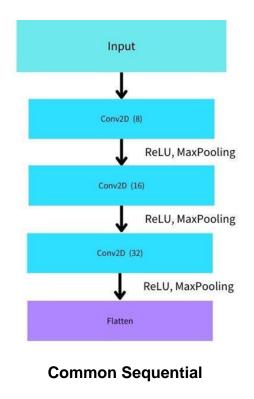


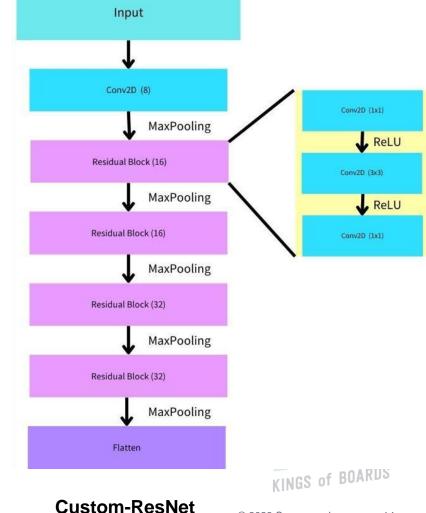
Custom ResNet vs. Common Sequential Model

ResNet significantly improved the accuracy

with acceptable numbers of parameters.

| | Sequential | Custom-ResNet |
|---------------------|------------|---------------|
| Total Parameters | ~50K | ~70K |
| Tflite Accuracy | 0.53 | 0.7 |
| | | |





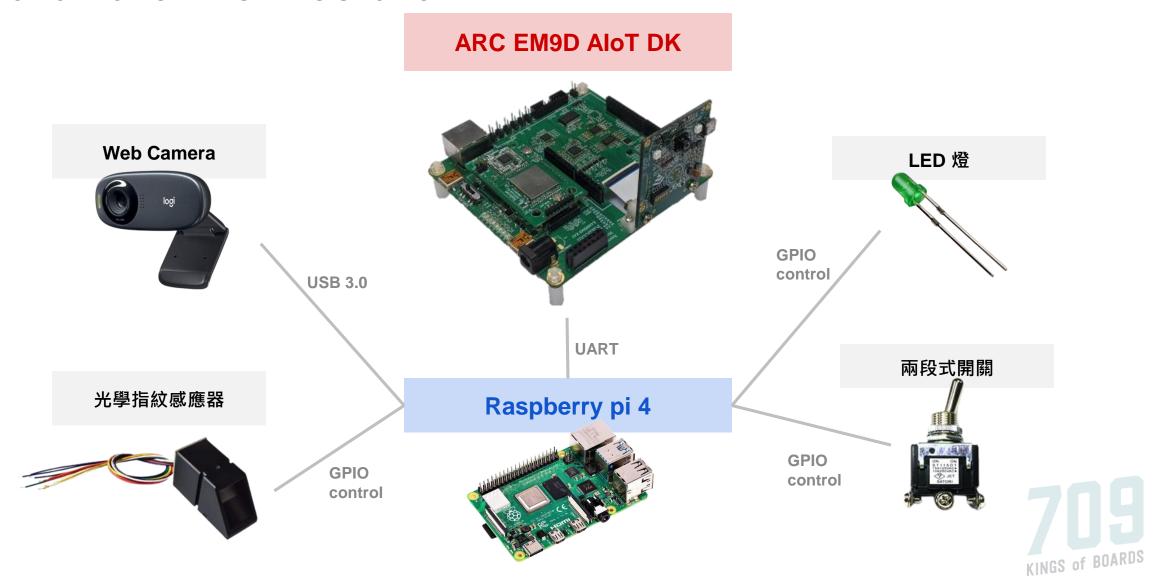
Hardware Compatibility with EM9D processor

- Low Latency Three-Stage Pipeline: Residual blocks can be efficiently
 executed in a low latency three-stage pipeline due to its relatively shallow
 depth.
- Low Power Consumption: ResNet models, especially those with fewer parameters, can be computationally efficient, which aligns well with the processor's goal of achieving low power consumption.

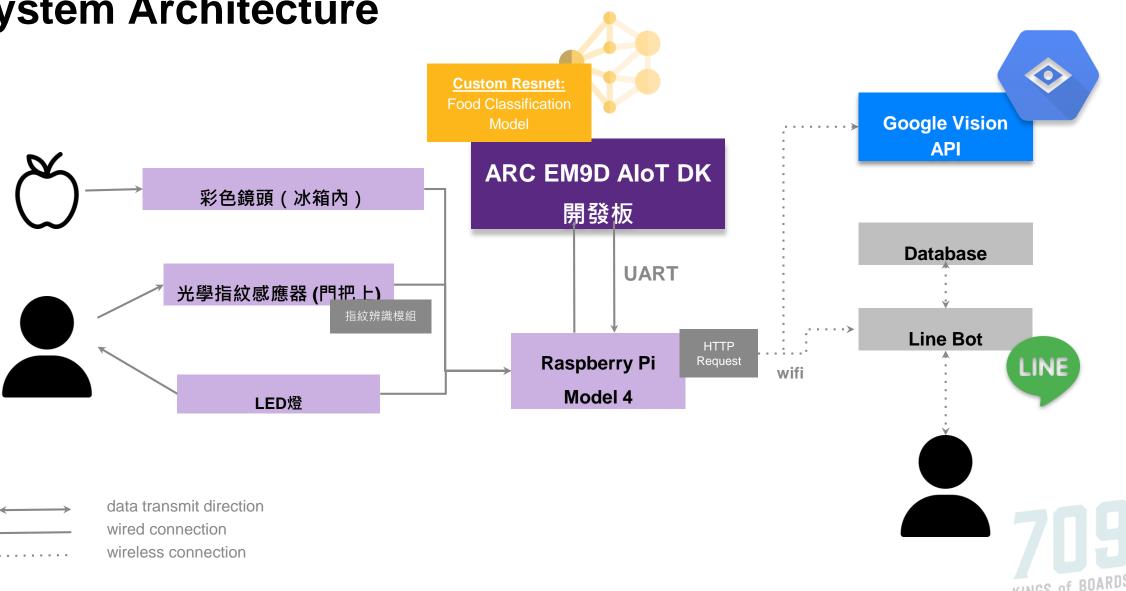
source: ARC EM9D and EM11D RISC + DSP Processors datasheet



Hardware Architecture



System Architecture



Hardware Spec

- 1. ARC EM9D AloT DK 開發板
- 2. 光學指紋感應器:AS608 Optical Fingerprint Reader Sensor Module
- 3. 食品影像辨識鏡頭: Logitech c310 webcam
- 4. Raspberry Pi 4



Software Spec

- 1. 食品辨識模型
 - CV CNN model : Self-Develop ResNet Model
 - Dataset : Fruits360
- 2. 指紋辨識模型
- 3. User Interface: Line Official
- 4. Database : MySQL Lite





Benefit of FoodieFortress

解決共用冰箱問題

食物不會被亂拿、 不再有過剩食品佔據 空間。



設計Custom Restnet Model

Multi-class的 food classification model,符合 Em9d處理器特性







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