



FOOD
STASH

Transforming Surplus into Sustenance

SAP x Food Stash | youCode 2025





The Team - CodeRed



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Background

Problem

Average food insecurity rate in Canada:

47%





Problem Statement

“While 1.3 billion tons of food is wasted annually, 828 million people go hungry due to broken donation systems plagued by high spoilage rates, unreliable volunteer networks, and poor communication between donors and recipients. This disconnect wastes perfectly edible food, generates harmful methane emissions, and leaves nearby shelters unable to consistently serve those in need”



Background

Challenge

- About 1 in 4 people in Canada are food insecure
- If food wastage was a 'country' it would be 3rd largest contributor of greenhouse gases,



Background

Goal

- Create a centralized platform to connect donors, volunteers, and shelters

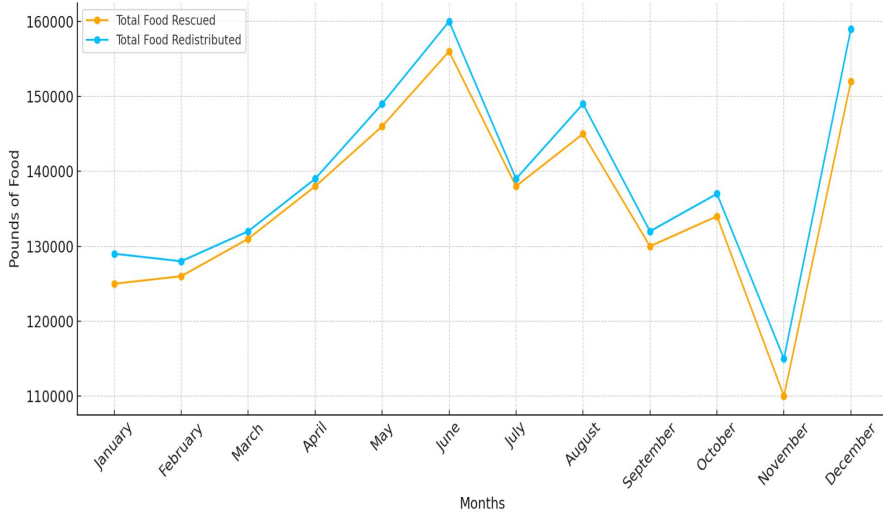
Solution

- Reduce spoilage by 15% in 6 months
- Match food types with community needs

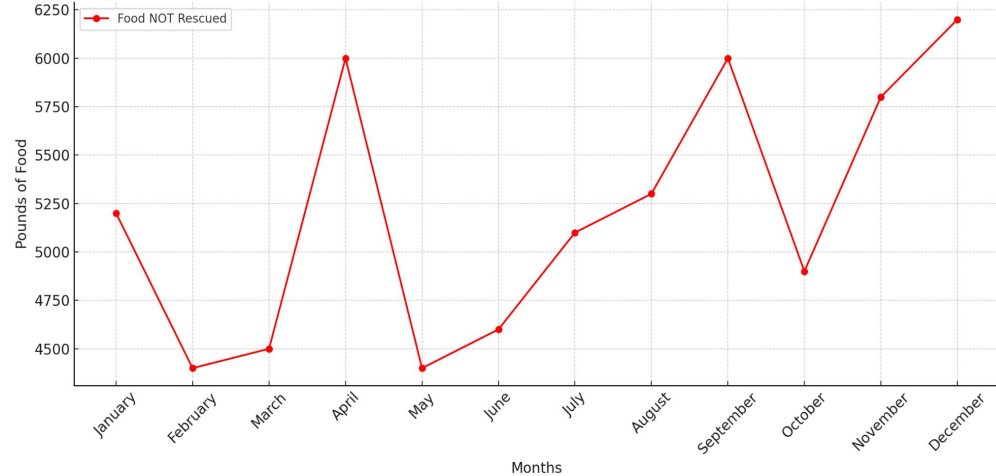


Month wise analysis of food rescued vs not rescued

Total Food Rescued vs Redistributed



Food NOT Rescued





Data Source



Food Stash 2024 data:

23% of avoidable food waste is caused by best before from processing to purchase dates . \$58B The total value of avoidable food waste every year.

Total Rescued

1,675,329 lbs

Total Not Rescued

62,266 lbs

Avg Rescued/Month

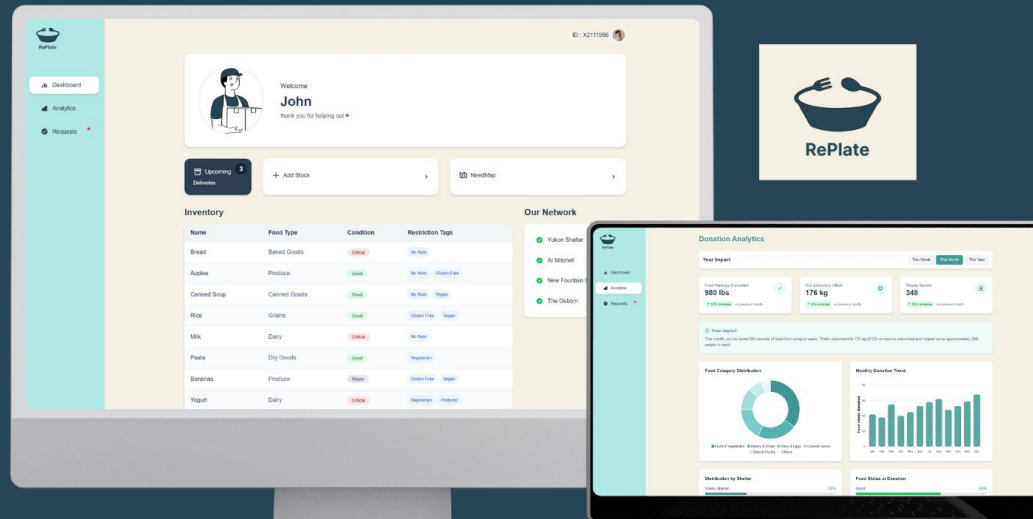
139,610 lbs

Spoilage Rate

3.6%

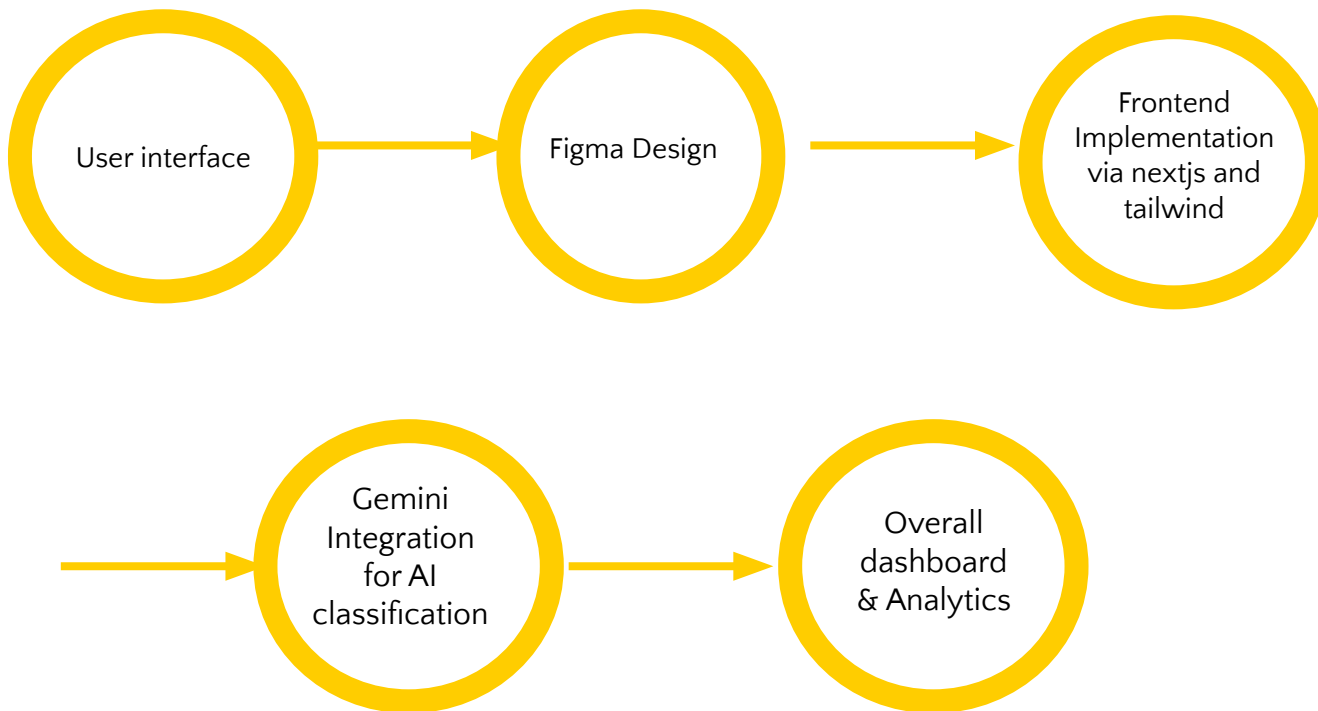


Website Demonstration



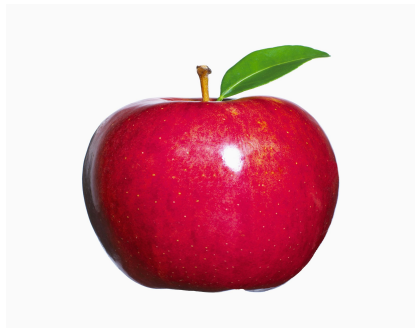


Model Pipeline





Testing Images Used






Method of Classification

- Fine Tuned & Tested 3 different models using custom images (for each fruit)
 - EfficientNetB0
 - MobileNetV2
 - ResNet50
- Models compared
 - For Banana: **Best Model** Taken (MobileNetV2) due to significantly **higher accuracy** (as shown in earlier result)
 - For Apple: **Average** of result of all 3 models displayed
- Further tuning image classification might be needed to **improve accuracy** (more images = higher accuracy).





ML Model Predictions - Bananas & Apples


 Prediction for 'bananaExpired.jpg'

Category	Probability (%)
Good	14.64%
Risky	10.30%
Expired	75.06%

Confident prediction: Expired (75.06%)


1/1  0s 25ms/step


 Classification Results


 Prediction for 'bananaGood.jpg'

Category	Probability (%)
Good	90.32%
Risky	8.88%
Expired	0.80%


Confident prediction: Good (90.32%)

1/1  0s 27ms/step



 Classification Results


 Prediction for 'bananaRisky.jpg'

Category	Probability (%)
Good	14.11%
Risky	47.55%
Expired	38.34%



 Average Predictions for 'appleExpired.jpg':


Category	Probability
Good	0.07%
Risky	22.54%
Expired	77.39%

 Prediction: EXPIRED (77.39%)  Moderate



 Average Predictions for 'appleGood.jpg':

Category	Probability
Good	97.12%
Risky	2.83%
Expired	0.05%

 Prediction: GOOD (97.12%)  Very Confident

 Average Predictions for 'appleRisky.jpg':

Category	Probability
Good	9.84%
Risky	79.44%
Expired	10.69%

 Prediction: RISKY (79.44%)  Moderate



Future Improvements

Expanding Our Reach

We're focused on expanding Re-Plate by onboarding local businesses and shelters in pilot areas and using their feedback to fine-tune the platform.

Smarter, Simpler Tools

a mobile app, geolocation matching, and volunteer routing to make donating surplus food faster, easier, and more efficient.

Boosting Impact

With in-app messaging and impact dashboards, we're strengthening connections and helping users see the real difference they're making in their communities



Thanks!

Any questions ?