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# Aquaponics: Deep Water Culture Systems

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# Overview

- In collaboration with Pure Reason AI, we constructed a classification model to identify nutritional deficiencies in lettuce
- Designed and trained convolutional neural networks
  - Using Python libraries Tensorflow and Keras
  - Fully nutritious (FN) or deficient in nitrogen (N), phosphorus (P), or potassium (K)



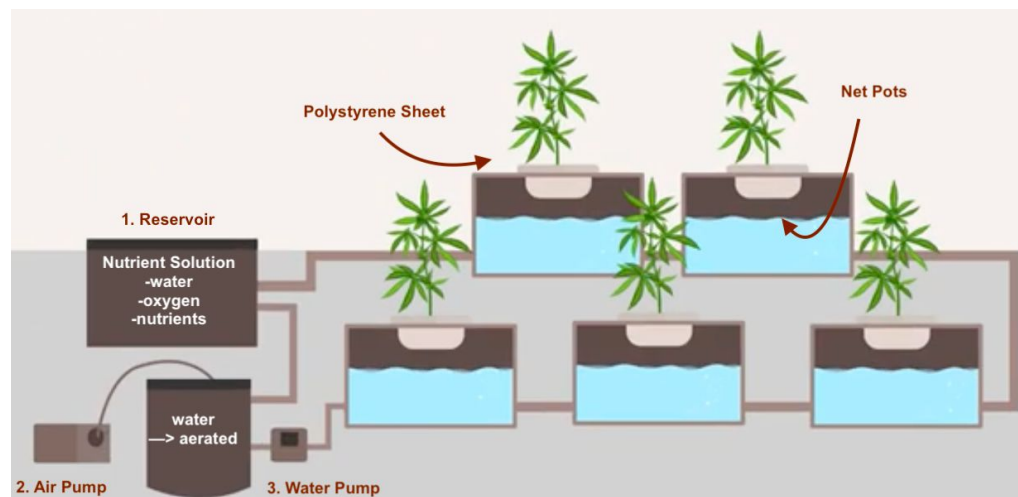
# Deep Water Culture

**Hydroponic system:** Soil-less agricultural system

**Deep Water Culture (DWC):** Hydroponic system where the root of the plant is submerged in an oxygenated, nutrient-rich water solution continuously

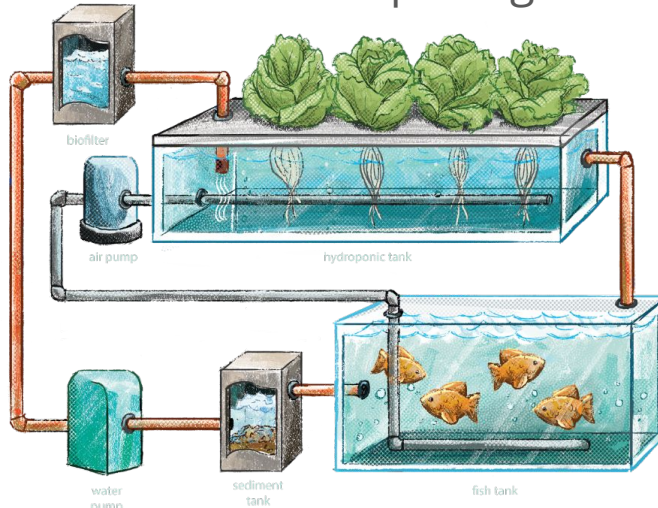
- Most efficient and simple system
  - 2-3x faster growth rate
    - ◆ More robust root formation
  - More water = more stable nutrient solution
    - ◆ Less monitoring
  - Basic structure
    - ◆ Less maintenance

- Adjust the nutrient levels in the solution
  - To maximize the quantity of nutritious plants per cycle



# DWC Aquaponic System

- **Aquaponics:** combination of aquaculture and hydroponics
- Our collaborators are developing a **DWC aquaponic system**
  - Nutrients from fish aid plant growth

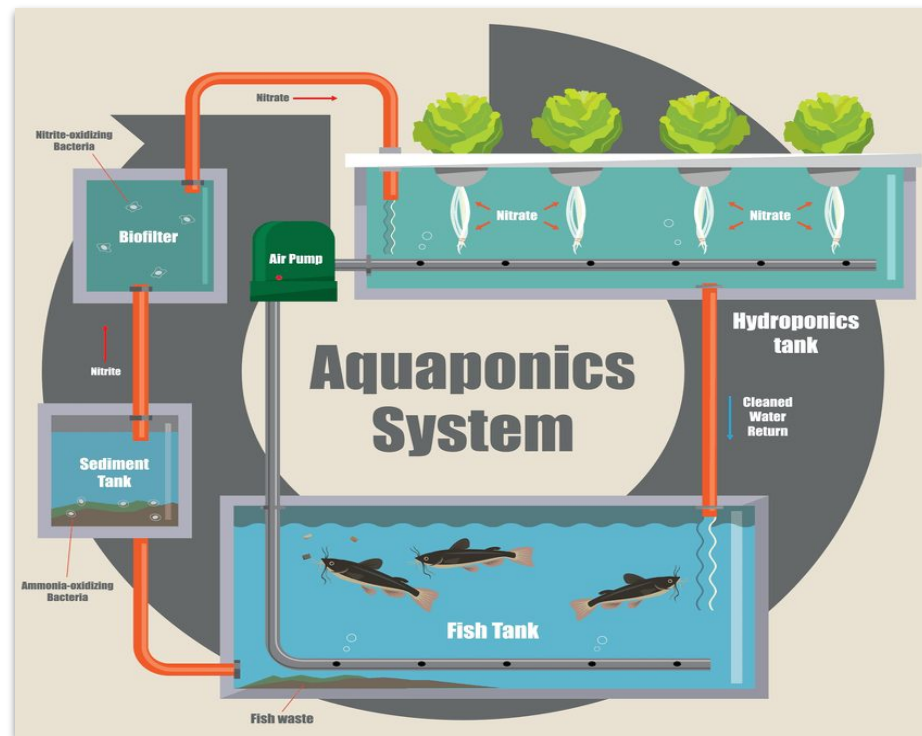


- Plants and fish must have similar temperature/ pH level needs
  - Catfish, bass, and tilapia are among the best to grow lettuce

	Temperature	pH Level
Lettuce	70-74°	5.8-6.2
Catfish	74-80°	5.5-7.0

# DWC Aquaponic System Process

1. Fish release nitrogen waste from their gills
  - In the form of ammonia
2. Water feeds into a sedimentation tank & biofilter
  - Sedimentation tank: removes solid waste
  - Biofilter: Beneficial bacteria performs nitrification
3. Water pumped into DWC tanks
4. Plants absorb the nitrate within the water
  - Preferred form of nitrogen when growing higher yields
5. Water becomes purified and re-circulates into the fish tank



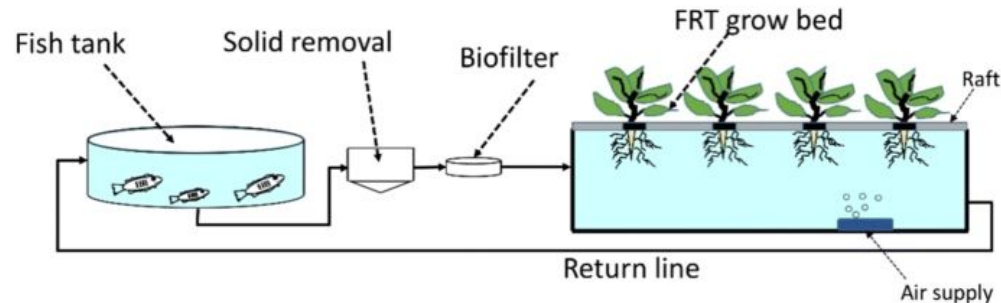
# DWC Aquaponic System Pros and Cons

## Advantages:

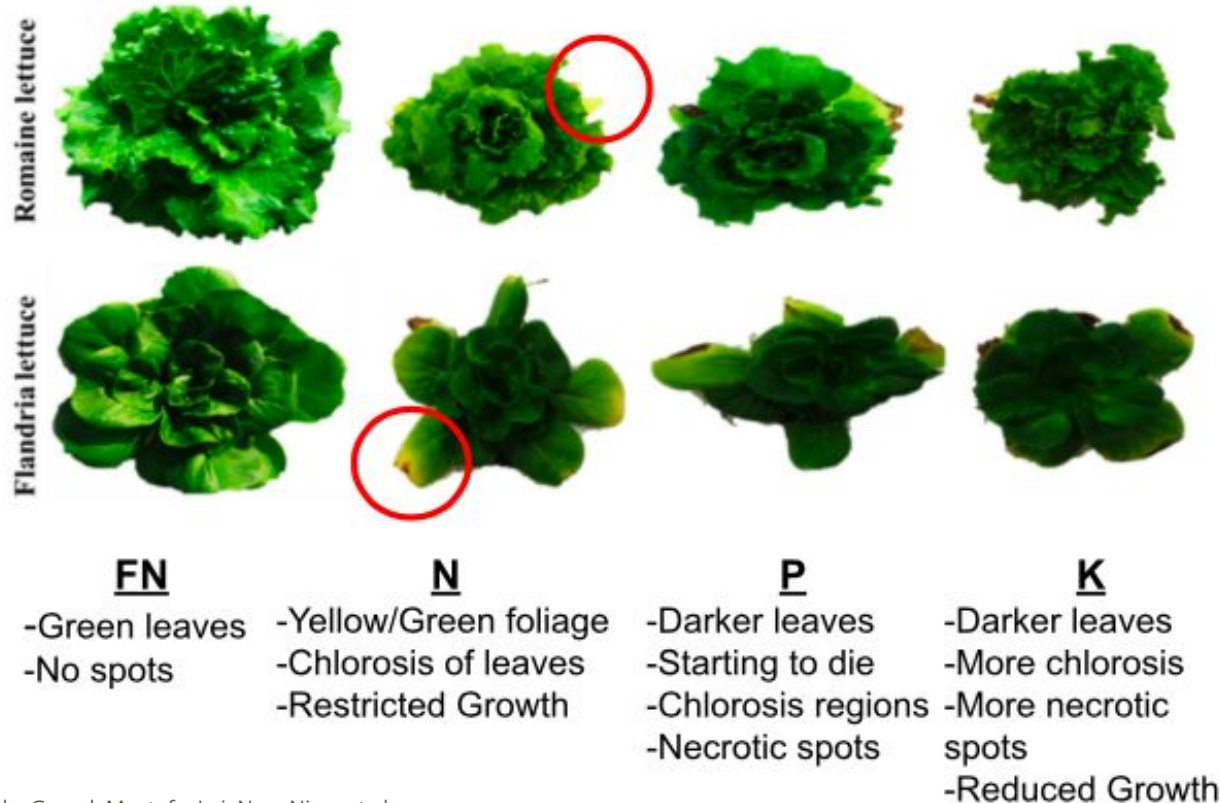
- Water-efficient
  - 90-99% less water
- No additional nutrient costs
  - Fish feed is the primary source of nutrient
- Eliminates soil-borne diseases
- Plants can survive up to 2 weeks w/o power
- 11% cheaper electric cost

## Disadvantages:

- Limited to small rooted plants
  - Leafy greens, lettuce, basil
- Higher Start-up/maintenance cost
  - Due to filtration system



# Identifying Nutritional Deficiencies



# Machine Learning Techniques

- Many projects have used machine learning techniques to help detect nutrient deficiencies in plants grown hydroponically
- The use of Convolutional Neural Networks have proven successful in detecting deficiencies by image classification of the plants
- Two projects have utilized CNN's to detect nutrient deficiencies:

## **Basil Plants:**

- 1. Used heat images**
- 2. 89.54% accuracy**
- 3. Challenges included wrongful detection and problems with shade and light in the images**

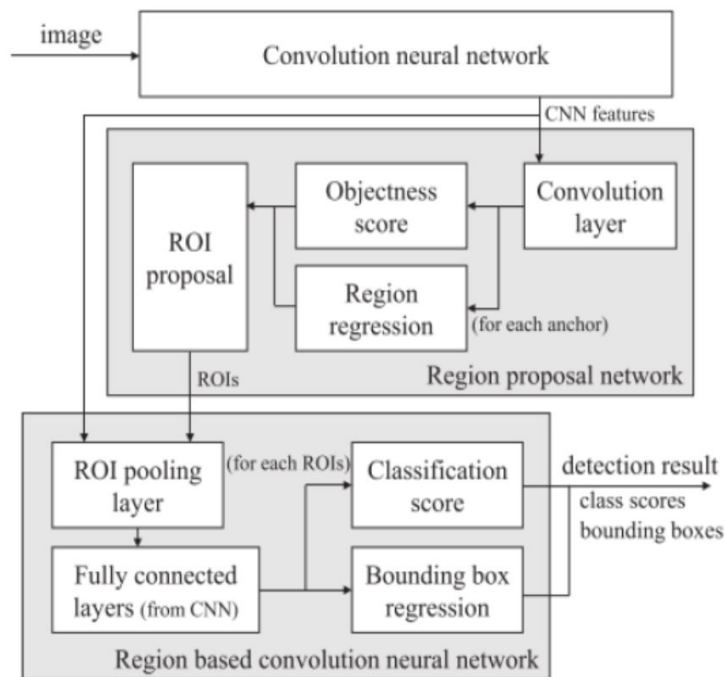
## **Lettuce Plants:**

- 1. Used a dataset of images**
- 2. Two algorithms ran**
- 3. Did not test accuracy of image classification and more work needed to be of practical use**



# Machine Learning Techniques Applied to Our Project

- Using similar ideas to the projects on the previous slide, our goal was to create a system to classify images of different lettuce plants
- Our project utilized CNN's and a collection of datasets to classify images based on the nutrient deficiencies found in the image



# Project Dataset

- Given to us by collaborators to train with
- Originally contained 300+ pictures
- Removed all corrupted images



Non-corrupted image



Corrupted image

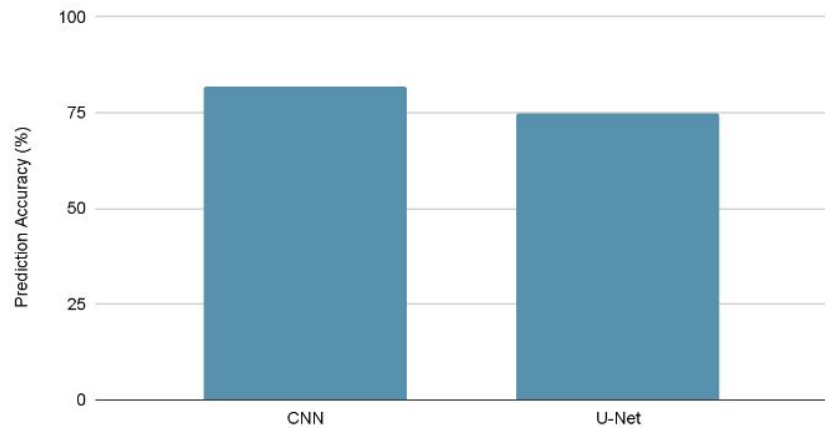
# Finished Dataset

Category	# of Images
FN	12
-K	72
-N	58
-P	66

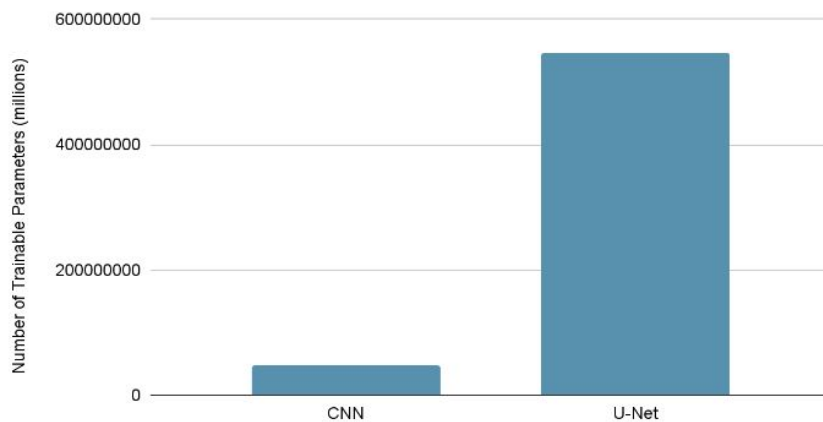
Total: 208 images

# Results

## Prediction Accuracy of NNs on Testing Data



## Number of Trainable Parameters

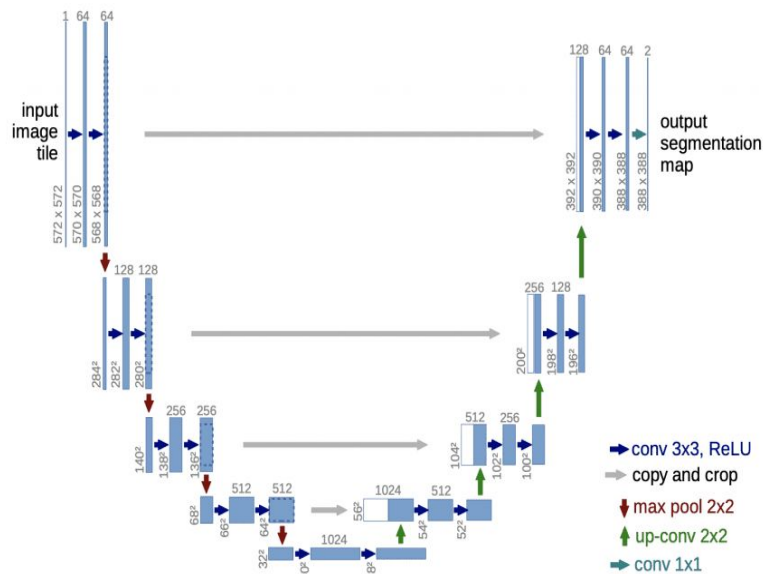
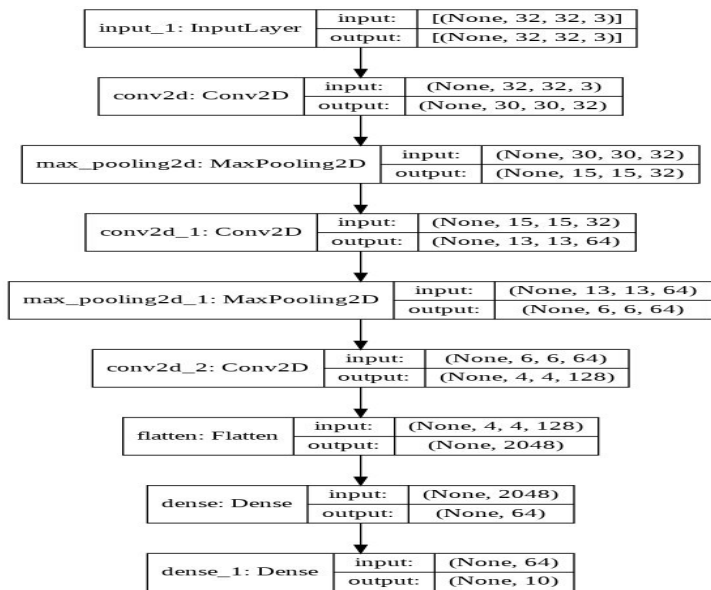




# Code

- Our task was to design a CNN's architecture, such as U-Nets, to help improve the accuracy of classifying images from their nutrient deficiencies with the given data set and determine the number of trainable parameters it used.
- We used TensorFlow and Keras to help developed our python code for the architectures.
- The two types of NN's we developed:
  - CNN (Sequential and Functional Models)
  - A more complicated type of CNN called a U-Net
- Then, we train and evaluate each model to determine the accuracy of prediction of the testing data set and the numberable of trainable parameters.

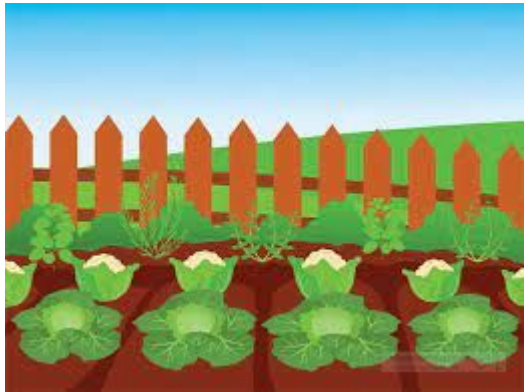
# CNN's Models



Layer diagrams for our CNN (left) and U-Net (right).

# Initial Investment/Cost for traditional

- Equipment Cost \$132,925
- Cost of Irrigation Per Acre \$1,216.50
- Average cost of an acre of farmland in Louisiana \$3,240
- Combined for 5 acres =  $5(4456.5) = \$22,282.50$
- Total Initial Cost \$155,207.50



# Traditional 5 Acre Farm Yearly Costs

Variable Cost: Roughly \$69,793

Fixed Cost: Roughly \$11,504

Total Yearly Cost: Roughly \$81,297





# Freight Farms Initial Investment/Cost

## Initial Purchase Cost

- \$149,000

## Shipping Fee from Boston MA

- \$7123.50 (\$4.50 per mile on average (1,583 Miles from here to Boston))
- \$1,200-\$3,000 for Crane to move the shipping container
- Optional Training

Initial investment - Roughly \$160,000-\$170,000



# Yearly Cost For Freight Farm

Supplies: Substrate, Nutrients and solutions: \$7,210

Equipment/Maintenance: Cleaning Supplies: \$1,500 Maintenance: \$2,000-\$3,000

CO2 Tank rental and Refill: \$2,000

Plastic Clamshell: \$40 per 100 and need 1000 every 4 weeks = \$5200

Energy usage: Roughly 190 kWh per day at \$0.1 per kWh = \$6935

Water usage: Minimal

Farm Software: \$2,400 a year

One freight farm requires on average 30 hours of labor per week

Yearly costs: \$26,245 + cost of 1560 hours of farming labor



# Product-Market Fit

- Geographic Constraints
  - Companies, such as Freight Farms, are mainly focused on selling to areas where cooling systems will be most cost effective
  - The extreme heat in places such as Louisiana lower overall demand, in turn forcing much higher prices
- Ideal Consumers
  - These companies have also identified wealthy agricultural companies as their main consumers.
  - Companies, comparable to the Kimbal Musk backed Square Roots, are the only ones who can pay the exuberant prices to buy these shipping container farms.
  - These companies are also in connection with much larger food markets, which allows for a more achievable return on their investment

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