CNN-powered Wheat Disease Detection



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GitHub
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The wheat industry

Worth **\$200 billion** USD – and growing.

Wheat Market Report Scope

Report Attribute	Details
Market size value in 2024	USD 202.01 billion
Revenue forecast in 2030	USD 270.88 billion
Growth rate	CAGR of 4.28% from 2024 to 2030

The world's food

Wheat is a worldwide source of food.

Wheat consumption today

With a massive production volume of 750 million tons per year, wheat is now a staple food for around 35% of the world's population. Of this production volume, nearly 70% is used for human consumption, with 20% reserved for animal feed and the rest used for other purposes such as biofuel production. To get to this point, the Green

Disease and pests

The cost is enormous.

NIFA AUTHORS

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Between 20% to 40% of global crop production is lost to pests annually. Each year, plant diseases cost the global economy around \$220 billion, and invasive insects around \$70 billion, according to the Food and Agriculture Organization of the United Nations. Weeds are another significant biotic constraint on global food production.

Disease and pests



Wheat rust



Aphid bugs

Billion dollar losses

+ Global food losses =

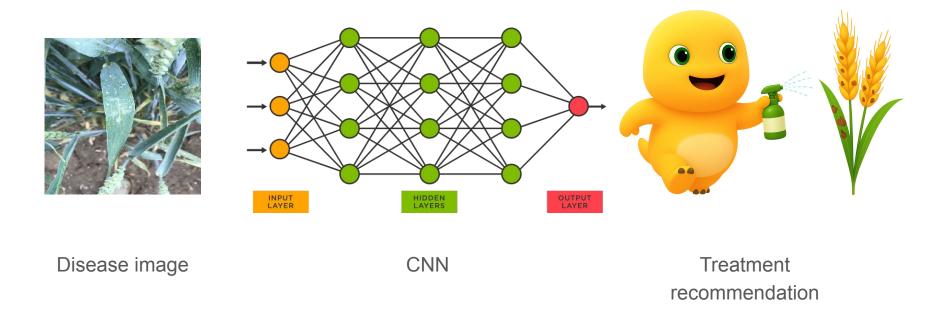
A very important problem to solve







Our project: wheat disease detection



Treatment recommendations

Based on model output → lookup corresponding treatment in table.

	Disease	Recommended treatments	Source links
0	Aphid	Imidacloprid seed treatment; Foliar insecticid	https://onlinelibrary.wiley.com/doi/full/10.10
1	Black Rust	Triazole fungicides (e.g., tebuconazole); Stro	https://www.cropscience.bayer.us/articles/cp/w
2	Blast	${\it Tebuconazole + Pyraclostrobin; Azoxystrobin +}$	https://ageconsearch.umn.edu/record/312377 l h
3	Brown Rust	Tebuconazole; Azoxystrobin + Tebuconazole mixt	https://extension.uga.edu/publications/detail
4	Common Root Rot	Seed treatment fungicides (fludioxonil, pyracl	https://extensionpubs.unl.edu/publication/g199
5	Fusarium Head Blight	Prothioconazole + Tebuconazole (Prosaro); Pydi	https://site.extension.uga.edu/turnerab/2022/0
6	Leaf Blight	Chlorothalonil; Mancozeb	https://hort.extension.wisc.edu/articles/home
7	Mildew	Triadimenol seed treatment; Azoxystrobin folia	https://plantpathology.ca.uky.edu/files/ppfs-a
8	Mite	Dimethoate; Pyrethroids (lambda-cyhalothrin)	https://extension.colostate.edu/topic-areas/in
9	Septoria	Propiconazole; Azoxystrobin	https://cropprotectionnetwork.org/encyclopedia
10	Smut	$\label{thm:proposed} \mbox{Tebuconazole seed treatment; Diffenoconazole se}$	https://eupdate.agronomy.ksu.edu/article_new/u
11	Stem fly	No effective chemical control; use cultural pr	https://efotg.sc.egov.usda.gov/references/publ
12	Tan spot	Tebuconazole; Trifloxystrobin	https://extensionpubs.unl.edu/publication/g429
13	Yellow Rust	Prothioconazole + Azoxystrobin (Trivapro); Teb	https://site.extension.uga.edu/turnerab/2022/0



Kaggle

Dataset can be found here.



KUSHAGRA3204 AND 2 COLLABORATORS · UPDATED A YEAR AGO







) :

Wheat Plant Diseases

Classification of various wheat plant diseases with almost 14,000+ images



Data Card

Code (8)

Discussion (3)

Suggestions (0)

About Dataset

Wheat Plant Diseases Dataset -

This dataset is designed to empower researchers and developers in creating robust machine learning models for classifying various wheat plant diseases. It offers a collection of high-resolution images showcasing real-world wheat diseases without the use of artificial augmentation techniques.

Usability ①

7.50

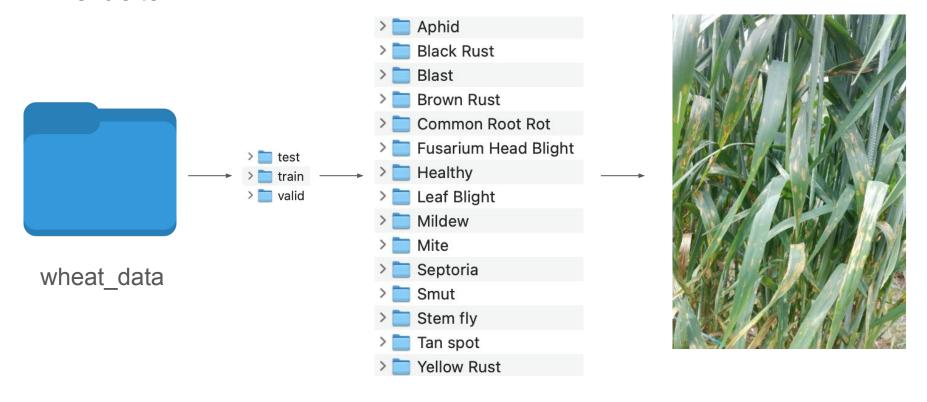
License

CC0: Public Domain

Expected update frequency

Annually

The data



14 disease classes, 1 healthy class.

We trained 3 CNNs and compared their performance

- ResNet18
- MobileNetV2
- Customized CNN

ResNet18

A good baseline for image recognition.

Changed output layer to match our 14 disease classes and 1 healthy class.

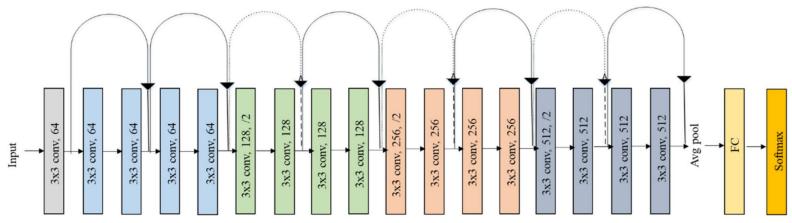
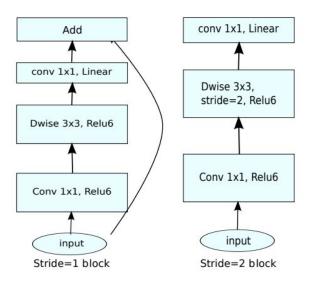


Fig. 2 Original ResNet-18 Architecture

MobileNetV2

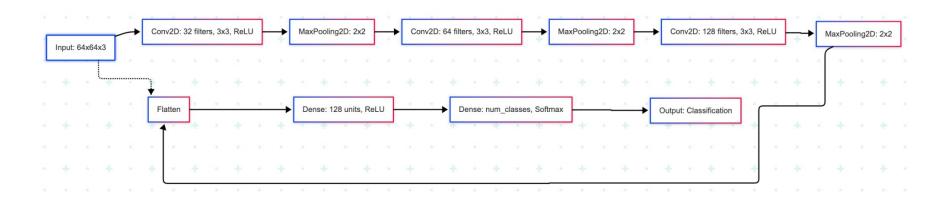
- Used for classification, object detection
- Reduce the number of parameters and computational cost.



Customized CNN

A homemade CNN from the ground up.

- Uses simple convolutions, no residual connections.
- Standard convolutions increase computational cost per layer.
- Overall relatively fast but less robust for complex patterns.



Training setups

Model	ResNet18	MobileNetV2	Customized CNN	
Layers	18	53	8	
Loss	Cross-Entropy	Cross-Entropy	Cross-Entropy	
Optimizer	Adam	Adam	Adam	
Learning Rate	0.0001	0.0001	0.001	
Epochs	20	10	30	
Early Stopping?	Yes	No	No	

Early results

Model	ResNet18	MobileNetV2	Customized CNN
Test set accuracy	92.40%	68.4%	90.27%
Average test class precision	0.952	0.677	0.917
Average test class recall	0.924	0.704	0.903
Average F1 score	0.908	0.681	0.885

Compare the performance of CNNs

- ResNet18 and the Customized CNN have a better performance on the classification.
- MobileNetV2 has a lower accuracy because it is lightweight with limited capacity, and has a weaker feature extraction ability.
- ResNet18 has more complex patterns and residual blocks, which captures richer and deeper representation.

Issues

Model is great at predicting disease – but struggles with healthy wheat

	Class	Accuracy (%)	Correct	Total
				100000000000000000000000000000000000000
0	Aphid	96.0	48	50
1	Black Rust	96.0	48	50
2	Blast	100.0	50	50
3	Brown Rust	100.0	50	50
4	Common Root Rot	100.0	50	50
5	Fusarium Head Blight	98.0	49	50
6	Healthy	10.0	5	50
7	Leaf Blight	96.0	48	50
8	Mildew	98.0	49	50
9	Mite	96.0	48	50
10	Septoria	100.0	50	50
11	Smut	100.0	50	50
12	Stem fly	100.0	50	50
13	Tan spot	96.0	48	50
14	Yellow Rust	100.0	50	50



Healthy or Disease?

The difference between healthy and diseased wheat is not always clear.



Healthy



Mildew

Possible Solutions

- 1. Data augmentation
- 2. Class balancing
- Loss Reweighting (higher penalty for incorrect healthy classification)



Data augmentation, class rebalancing, custom loss results

It is difficult for image recognition models to detect an absence of disease!

Class	Accuracy (%)	Correct	Total	Class	Accuracy (%)	Correct	Total
Aphid	76.0	38	50	Aphid	66.0	33	50
Black Rust	56.0	28	50	Black Rust	46.0	23	50
Blast	88.0	44	50	Blast	96.0	48	50
Brown Rust	30.0	15	50	Brown Rust	24.0	12	50
Common Root Rot	90.0	45	50	Common Root Rot	78.0	39	50
Fusarium Head Blight	62.0	31	50	Fusarium Head Blight	42.0	21	50
Healthy	6.0	3	50	Healthy	6.0	3	50
Leaf Blight	70.0	35	50	Leaf Blight	48.0	24	50
Mildew	78.0	39	50	Mildew	54.0	27	50
Mite	60.0	30	50	Mite	30.0	15	50
Septoria	92.0	46	50	Septoria	84.0	42	50
Smut	92.0	46	50	Smut	78.0	39	50
Stem Fly	78.0	39	50	Stem Fly	74.0	37	50
Tan Spot	48.0	24	50	Tan Spot	40.0	20	50
Yellow Rust	100.0	50	50	Yellow Rust	98.0	49	50

Original

New

Further research and final thoughts

- More images for healthy wheat
- Powerful architecture ex.
 EfficientNetB3
- Add attention mechanism, focus on disease patterns



Thank you!