



ಭಾರತೀಯ ಮಾಹಿತಿ ತಂತ್ರಜ್ಞಾನ ಸಂಸ್ಥೆ ರಾಯಚೂರು
भಾರತೀಯ ಸೂಚನಾ ಪ್ರौದ್ಯೋಗಿಕಿ ಸಂಸ್ಥಾನ ರಾಯಚೂರು
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Automated Cyclone Eye Localization: A Comparative Study

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INTRODUCTION

- Tropical cyclones derive their primary energy from the evaporation of warm ocean waters .
- The storm is caused by the Coriolis Effect.
- Eyewall surrounds the calm, low-pressure "eye" seen in your satellite imagery.

USD 1.4 trillion

Economic losses caused by tropical cyclones during the 50-year period since 1970.

200-500 km

The diameter of a tropical cyclone is typically around 200 to 500 km, but can reach up to 1000 km.

43 deaths

and USD 78M in damages caused by tropical cyclones and related hazards every day in the 50-year period since 1970.

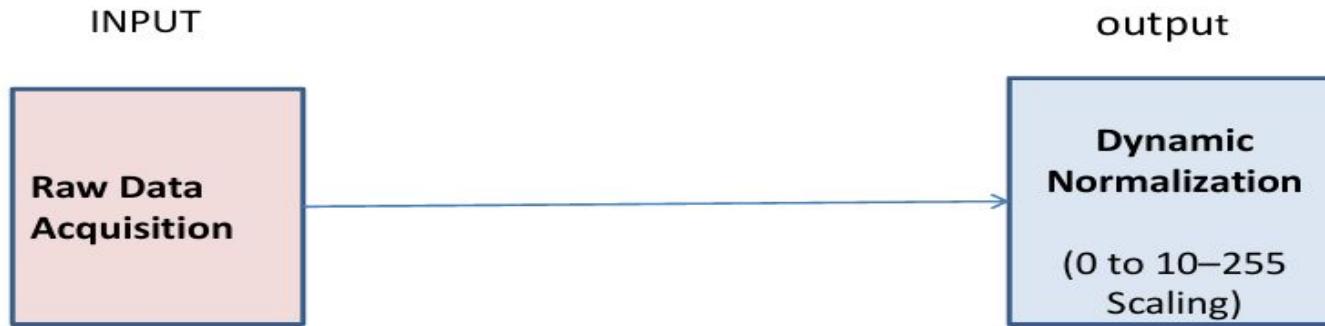
200%

Percentage of which the population of tropical cyclone-prone regions has increased.

AIM & OBJECTIVES

- Develop a robust HYBRID localization framework for cyclone eye detection
- Implement a multi-stage pipeline using Dynamic Normalization and hybrid segmentation.
- Conduct a comparative study between standalone DNN and hybrid-preprocessed models.
- Validate the proposed framework using unseen satellite datasets to ensure reliability.

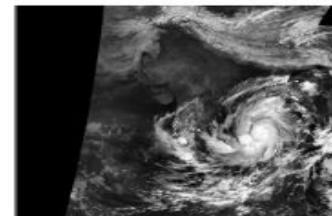
METHODOLOGY



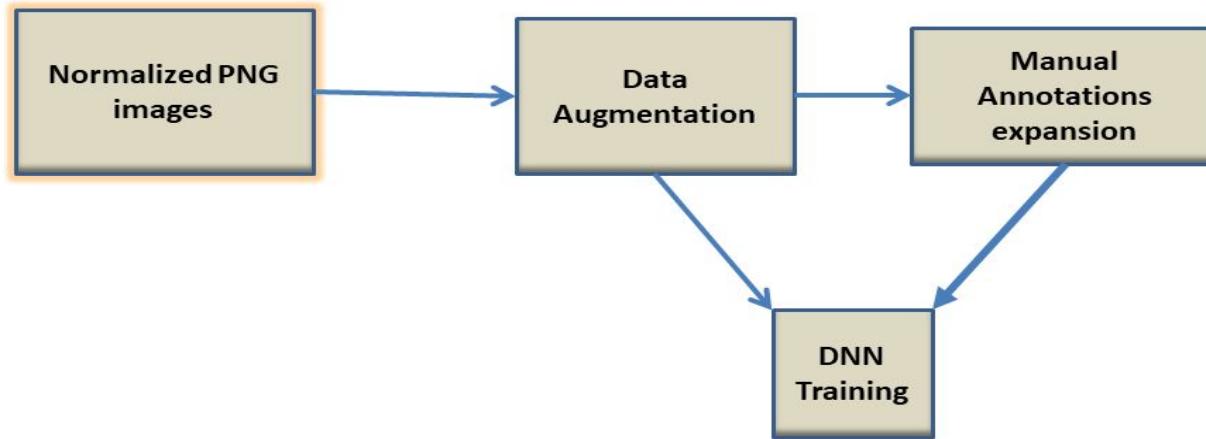
Satellite TIFF
images with
temperature
values

Image Analysis for: 20141009-194030Z-212000Z

- ▶ Minimum Intensity: 0
- ▶ Maximum Intensity: 255
- ▶ Zero Count (Black Pixels): 3,250,292
- ▶ Background Coverage: 16.54%

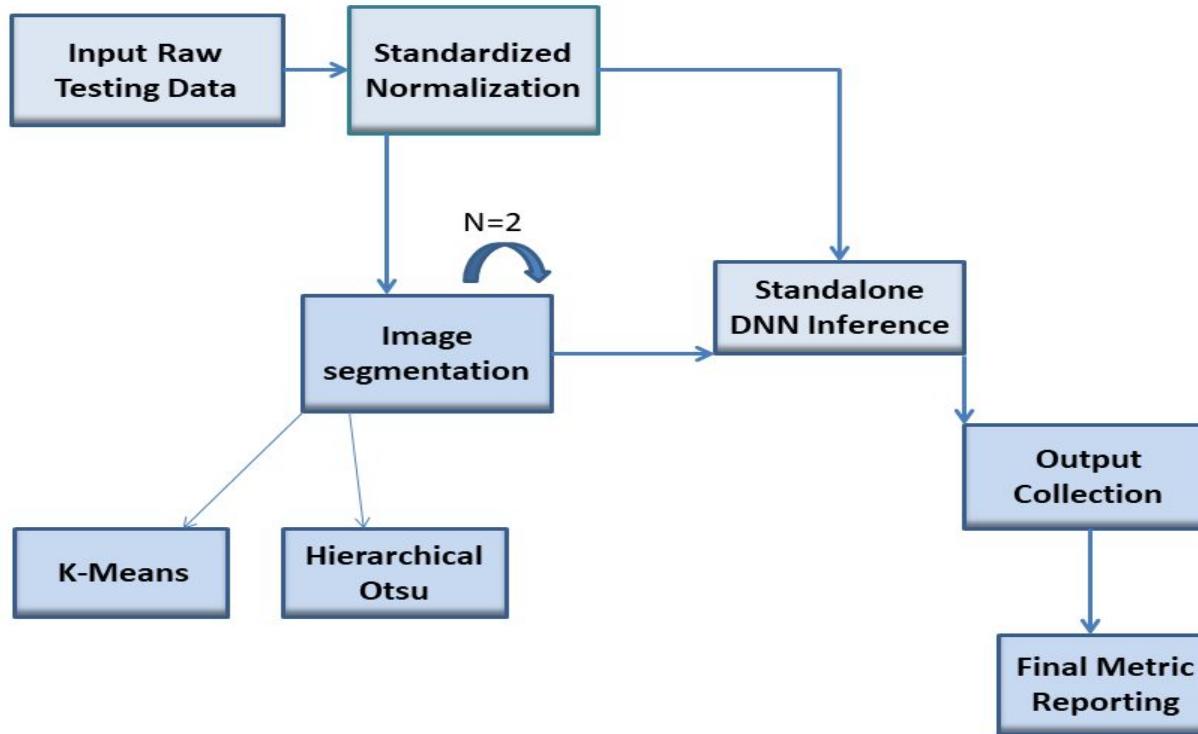


TRAINING PIPELINE



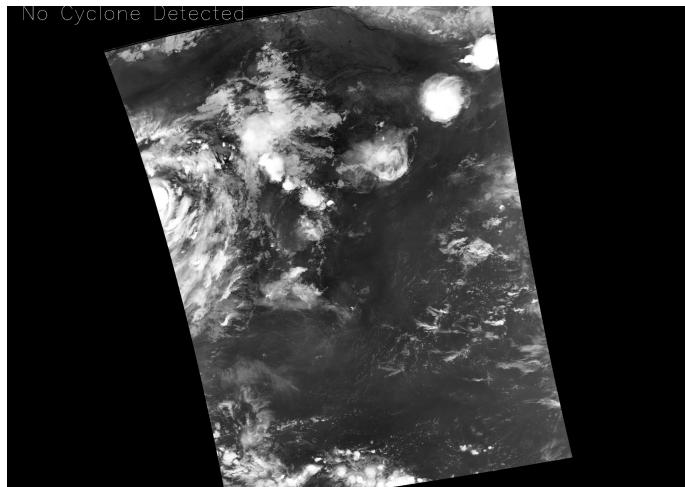
After the annotations were generated
Then the model will be trained along
With the images.

METHODOLOGICAL EVALUATION



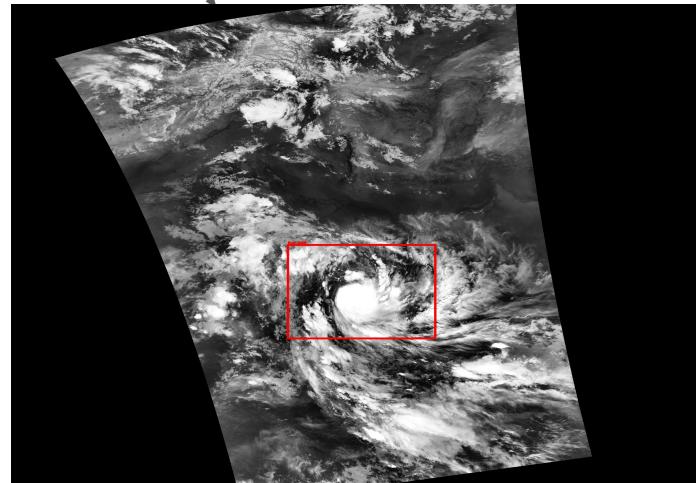
RESULTS

If no



Inference on unseen data

If yes



RESULTS

Method	IOU(0-1) (mean \pm SD)	CDE (Mean \pm SD)	mAP@50
DNN(YOLOV8)	0.469 ± 0.094	990.8 ± 969.7	0.533
DNN + Hierarchical OTSU	0.569 ± 0.168	2227.7 ± 1593.5	0.500
DNN + k-means	0.648 ± 0.247	1400.7 ± 2220.1	0.700

- DNN (YOLOv8) baseline is satisfactory
- DNN + k-means architecture has performed the best among all tested variants

CONCLUSION

- This research successfully demonstrates that a Hybrid Deep Learning Framework significantly outperforms standalone deep neural network.
- in the localization of tropical cyclone eyes. By integrating K-Means(15.4% improvement) and Hierarchical Otsu preprocessing, we effectively eliminated land-based thermal interference, leading to more robust feature extraction by the YOLOv8 model.

REFERENCES

- **Jocher, G., et al. (2023).** *Ultralytics YOLOv8*. (Software framework used for training the **best.pt** weights and performing cross-validation).
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- **Lloyd, S. (1982).** *Least Squares Quantization in PCM*. IEEE Transactions on Information Theory.

Thank you..