

An AI-based 3D Bat Movement Tracking System at Wind Energy Facilities using Multi-Thermal Video Cameras



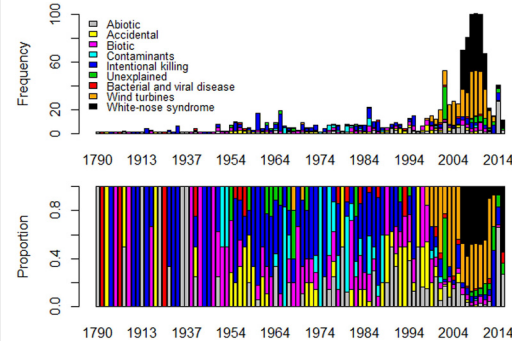
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Motivation

Growing Concerns of Increased Bat Mortalities

Motivation for building this kind of tool comes from the **dead bats**. Bats are important for ecosystems, providing services like pest control and pollination. However, starting around 2000s, the global increase of industrial wind-power facilities contributed significantly in bat mortality causes.



O'Shea, T.J., Cryan, P.M., Hayman, D.T.S., Plowright, R.K. and Streicker, D.G. (2016). Multiple mortality events in bats. Mammal Review, 46: 175-190. <https://doi.org/10.1111/mam.12064>

Studies of bat behavior and interactions with turbines have occurred for nearly two decades and yet, **little is still known about their behavior and interactions with turbines** during their active period, limiting the research community's understanding of the risk. The capability to **closely monitor the real-time behavioral states or mere presence of bats flying in the rotor-swept airspaces around wind turbines** during nights is crucial. It facilitates expedited data retrieval, providing valuable insights to enhance our understanding of the factors driving risks to bats from wind energy.

Cost-effective Strategy to Monitor Bat Activities

Characterizing patterns of behavior that increase the risk to bats at wind turbines while collecting real-time environmental conditions associated with such risk requires **consistent, long-term, easy-to-implement, and broadly scalable observation and monitoring methods**. Thermal-infrared imaging (hereafter 'thermal video cameras') record video of bats flying near wind turbines emerges as a promising technology for investigating bat behavior, offering continuous monitoring capabilities around wind turbines.

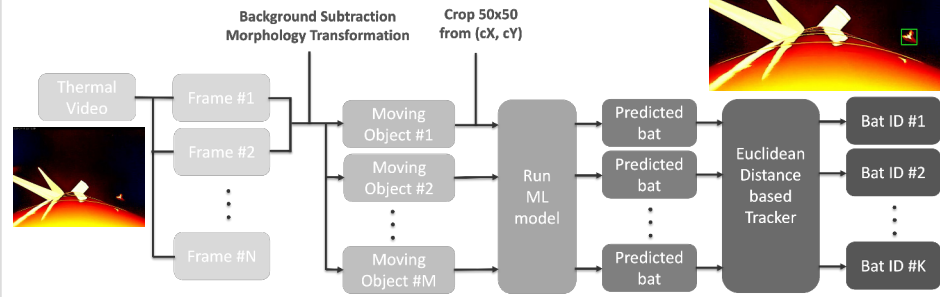
However, the sheer volume of data generated by continuously operating thermal-imaging systems surpasses **human capacities for manual review**. To leverage the potential of real-time thermal-imaging methodologies in quantifying nocturnal bat activities at wind turbines, we have pioneered **3D computer vision techniques within a deep learning framework**. This innovation enables the automatic detection and classification of bats, birds, and insects in thermal-imaging videos captured at wind turbine sites, facilitating efficient and accurate data analysis for enhanced understanding and mitigation of bat-wind turbine interactions.



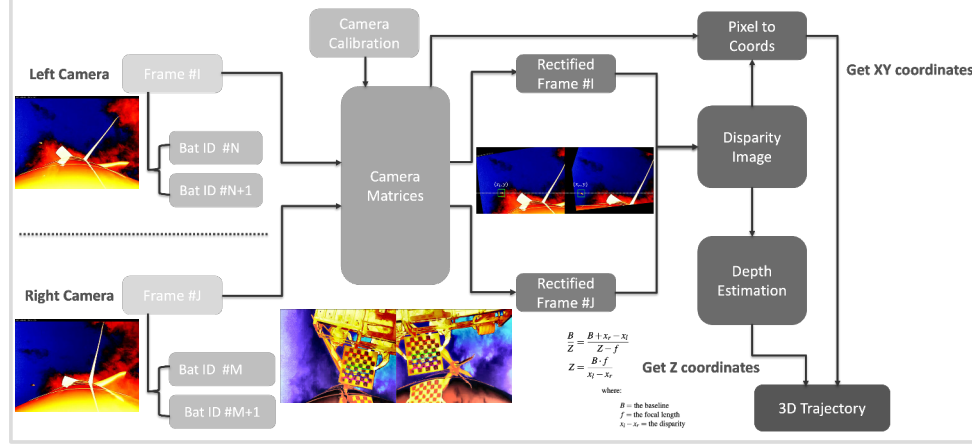
Three thermal cameras (AXIS Q1941-E Thermal Network Camera). Each sensor was mounted 70.25 inches off the ground/up the turbine and were 37.5 inches apart, pointing up at the blade. All recordings were collected with matching timestamps using a GPS module.

Methodology

2D Bat Movement Tracking Workflow

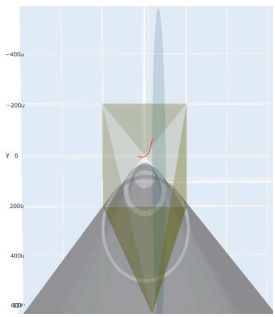


3D Bat Movement Tracking Workflow



Conclusion

- Moved the research community and industry closer to fully automated wildlife behavioral analysis and real time reporting to help expedite the research
- Built multiclassification model with the ability to distinguish between bats (93.2%), birds (94.4%), insects (83.5%) and non-biological objects (99.1%)
- Cost effective thermal tracking system that
 - is computationally inexpensive with the ability to run in real time on a personal computer
 - can detect and classify animals moving in the field of view
 - can provide the 3D flight trajectories
 - can be easily retrained to enhance performance on other types of objects and applications
 - is entirely open source, allowing it to be accessed, maintained, understood, adapted, and improved upon by a broad range of end users. (<https://github.com/NREL/WEBAT>)
- Future works
 - Expand the monitoring system to include other type of sensors
 - Bat behavioral studies
 - Understand the interaction of bats with wind turbines



Bat Detection

Experimental Set up

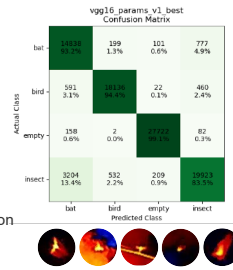
- GPU setting with Skynet from the Bioenergy Science Technology Directorate at NREL
- NVIDIA RTX3090 with Driver Version 470.57.0, CUDA, cuDNN, TensorFlow library

Dataset

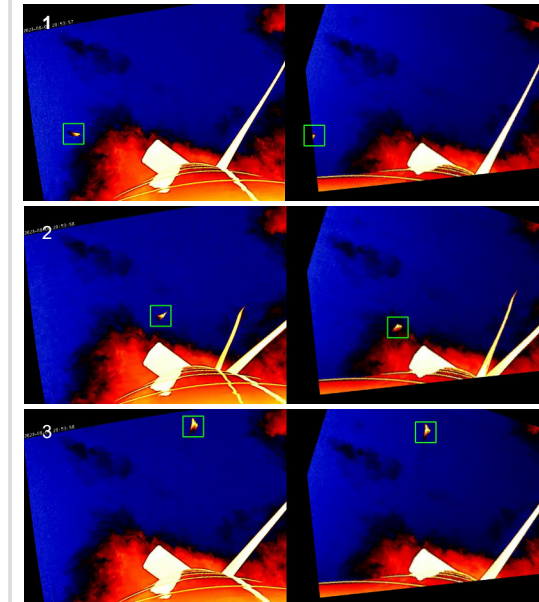
- 870,033 images** of unified 50x50 size
- 4 classes – bat, bird, insect, and non-biological object

Transfer Learning

- Backbone model option:
 - VGG, ResNet, Inception, Xception
- Hyperparameter choices:
 - # of freezing layers, # of classification layers with units, weight initialization, optimizer, learning rate, etc.
- Achieved **0.927124** for the averaged test accuracy :
 - bats(93.2%), birds(94.4%), insects(83.5%), non-biological objects(99.1%)



3D Bat Movement Tracking Results



Date	Time	Object Type	ID	X (cm)	Y (cm)	Z (cm)	Distance (cm)	Speed (cm/s)	Speed (mph)
8/8/23	20:52:38.767	bat	282	102.039	-136.199	3147.524	16.8578438	510.843752	11.4272493
8/8/23	20:52:38.800	bat	282	-85.3062	-134.804	3146.019	25.54073571	381.205011	8.52731323
8/8/23	20:52:38.867	bat	282	-60.9044	-127.832	3143.142	12.60691299	382.027664	8.54571546
8/8/23	20:52:38.900	bat	282	-49.0522	-129.227	3139.079	27.01399465	403.199502	9.01918178
8/8/23	20:52:39.000	bat	282	-14.8898	-132.715	3127.92	10.92925229	331.189463	7.4084979
8/8/23	20:52:39.034	bat	282	-3.73471	-133.41	3120.779	13.26325241	390.05659	8.72619137
8/8/23	20:52:39.200	bat	282	38.0968	-162.692	3112.309	51.75971537	311.805514	6.97489071