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Abstract

The paper presents the results of practical tests of an UAV aggregated with a thermal imaging camera for detecting animals in forested areas during the vegetation season (specifically in summer). The study focused on estimating the likelihood of detecting eurasian elk, red deer, european roe deer, and eurasian wild boar.

The UAV's flight altitude was 80 meters above ground level, providing a ground sampling distance of 7 cm. The tests were conducted between June 4 and September 26, 2022. A total of 21 flights were completed, covering a surveyed section approximately 110 kilometers long. The width of the inventoried transect was approximately 45 meters. Based on the total length of the route, the surveyed area was approximately 495 hectares, representing around 7% of the forested area in the Czarna Białostocka sub-district, where the research was conducted.

In conclusion, the results indicate that summer is not the optimal time for conducting animal inventory flights. This is evidenced by the fact that approximately 60% of all thermal signatures detected by the observer could not be attributed to specific species. The dense tree foliage during this season likely contributed to this challenge. It's also important to highlight that the forest habitats in the study area are highly fertile, featuring not only a canopy of trees but also thick undergrowth and second-growth forests, which can suppress the heat radiation emitted by animals. In less fertile areas, where understory vegetation is absent and the tree cover is primarily composed of coniferous trees (which have foliage year-round), the season of observation may have less impact on data quality.

Keywords: Animal detection, UAV, Hexacopter, Thermovision

1. Introduction

The range of technologies currently employed in nature conservation is remarkable, with many modern applications being adapted from other fields and tested in real-world environments. Among these, Unmanned Aerial Vehicles (UAVs) are becoming increasingly popular [15]. UAVs are now widely used across various sectors, including natural resource inventory.

Data collected using UAVs, when combined with specialized equipment, is often more accurate than that obtained through traditional ground-based surveys. UAV-based surveys also tend to be less stressful for animals compared to conventional methods [4, 22]. Even in ground-based inventories, monitoring multiple species presents challenges, as different animals typically require distinct monitoring techniques. It is crucial to consider the behavior of the species being surveyed [11]. Before undertaking a population survey, it is essential to establish both the purpose of the inventory and the manner in which the data will be applied [10]. In addition to direct methods, such as counting individual animals, indirect methods can also be employed to detect their presence [7, 8, 23]. These indirect methods do not require directly seeing or hearing the animal but involve identifying traces of their presence, such as tracks or droppings. The use of UAVs for estimating

animal populations is generally considered a direct method, as the goal is often to detect individual animals. However, this is not always the case.

UAVs equipped with cameras operating outside the visible light spectrum can enhance the effectiveness of wildlife surveys [2]. Thermal sensors, for instance, detect temperature differences between the target and its surroundings [16]. The use of thermal imaging cameras (TIR) on drones is particularly effective for wildlife detection, as these cameras capture heat emitted by animals, unlike traditional cameras, which rely on visible light reflected from objects. The key factor in utilizing thermal imaging is the animal's body temperature being higher than its environment (thermal contrast). While atmospheric distortions may occur when measuring temperatures from UAVs, these distortions are typically negligible at low altitudes [1], which is significant when using UAVs equipped with thermal imaging.

Efforts are underway to automate the analysis of images for animal detection. Research indicates that automatic detection using RGB sensors on UAVs is highly feasible for large species inhabiting open, homogeneous environments with minimal vegetation or topographic variation. Infrared sensors, integrated with multi-rotor UAVs, are more effective for detecting smaller, elusive species in complex habitats. Corcoran et al. provide valuable insights into this subject [6].

During the summer season, the presence of foliage and undergrowth can limit the detection of animals [21]. The readability of thermograms is also affected by wind, which not only impedes UAV flights but also lowers the animal's body temperature by reducing thermal contrast. This makes it difficult to detect animal signatures in thermal images and videos [20]. A significant challenge in using UAVs for wildlife observation is that animals often seek shelter under vegetation [12, 18], which can block emitted radiation. However, precise temperature readings of the entire animal are not necessary for detection; it is sufficient for the animal to appear as a hotspot against the cooler background [5].

UAV imagery is not the only useful tool for animal detection. Camera traps also play an important role in this process [13, 14]. The automation of species identification using camera traps shows promising results [3]. With camera traps (whether ground-based or tree-mounted), individual animals are more likely to be identified using deep learning or YOLO algorithms, as camera traps have the advantage of capturing the full figure of the animal, unlike UAVs, which typically provide an overhead view. Consequently, automatic species identification is expected to be more challenging with UAV images than with photos from camera traps. There is significant potential in automated object detection, though no current algorithm is yet fully applicable. For instance, research by Chao Mou et al. [17] on real-time animal detection yielded encouraging results, but theirs study (according to the article) was limited to species found in open or sparsely wooded areas, so the method may not be applicable in other environments.

The proportion of fertile habitats in the area is also relevant, as these habitats typically support dense canopies and substantial ground cover. Additionally, the structure of forest stands in such areas is often more complex, with dense upper-storey canopies and significant understorey vegetation. An interesting study by Pagacz et al. examined the degree of ground cover by tree and shrub canopies during the winter period [19].

- 1.1 Definitions, acronyms and abbreviations
- Definitions, acronyms and abbreviations
- AGL = Above Ground Level;
- GSD = Ground Sampling Distance;
- RGB = Red-Green-Blue;
- UAV = Unmanned Aerial Vehicle.

2. Research site and main research interest

UAV tests with a thermal imaging camera were conducted in the Czarna Bialostocka Forest District during the summer of 2022. These tests of the unmanned aerial vehicle are a continuation of the

research previously described by the authors [9]. For the 2022 summer season discussed in this article, the flight methodology was modified compared to earlier seasons

2.1 Purpose

The aim of the study was to assess the usefulness of thermal imaging camera data collected from UAVs during the summer for detecting wildlife in forested areas. The research focused on the following species: red deer, eurasian elk, roe deer, and wild boar.

2.2 Research area – Czarna Bialostocka Forest sub-district

The Czarna Białostocka area was selected as the research site. The Czarna Białostocka precinct covers approximately 6,980 hectares. Figure 1 shows selected flight transects within the Czarna Białostocka precinct, where the flights were carried out.



Figure 1 – Selected flight transects (marked in blue) in the Czarna Białostocka precinct of the Czarna Białostocka Forest District, where research flights were conducted

2.3 Equipment

The selected equipment included the following:

a) E20Tvx, Yuneec:

Justification: The camera's low viewing angle (33° x 26.6°) allows for the acquisition of high-quality images due to the smaller Ground Sample Distance (GSD) compared to cameras with a larger

viewing angle. Additionally, it offers high sensitivity and a high-resolution thermal sensor matrix (650 x 512). Nevertheless, it should be noted that there are cameras of much higher quality then the one used in this study. However, some of mentioned cameras may by used only under military restricions.

b) Hexacopter Yuneec H520E:

Justification: This UAV was chosen for its compatibility with the selected camera.

2.4 Flight conditions

The flights were conducted during the summer season, from June 4 to September 26. Additional flight conditions included:

- one battery was used per mission, with no battery changes during a single mission,
- the centres of adjacent transects were spaced 120–150 metres apart,
- flight altitude was maintained at 80 metres above ground level (AGL),
- horizontal flight speed was set at 10 m/s,

• the take-off and landing site for the multi-rotor was located at least 200–300 metres from the surveyed area to minimise the impact of noise on the animals being inventoried,

• the survey area was inventoried starting from the point nearest to the operator and progressing through the surveyed area.

3. Findings

A total of 21 flights were completed, covering a surveyed section approximately 110 kilometres in length (Table 1). Based on the horizontal field of view of the thermal imaging camera (33°) and the flight altitude (80 metres AGL), the width of the inventoried transect was approximately 45 metres. Given the total length of the route, the surveyed area covered approximately 495 hectares, representing around 7% of the forested area in the Czarna Białostocka sub-district.

As a result of all 21 UAV flights, only one red deer was detected. No moose were identified during the summer flights. Based on tests conducted in previous seasons, it can by stated that the quantity of animals detected during summer perioid is low [9]. Six european roe deer and seven eurasian wild boar were detected. However, it is important to note that previous tests have shown that the equipment used did not provide reliable identification of these species (european roe deer and seven eurasian wild boar) [9]. A total of 21 thermal signatures were recorded but not classified. Approximately 60% of all thermal signatures detected by the observer could not be attributed to specific species. The main reason for the inability to identify the species was dense vegetation. A summary of the number of recorded individuals is presented in Table 2.

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Table 1 Summary of flights conducted during the summer of 2022

Date	european roe deer	red deer	eurasian wild boar	eurasian elk	not marked	Sum	Approximate flight location		Inventoryed area (surface)
	number of individuals						numb divis	ers of ions	ha
4 \/I						0	22	23	23,58
4 VI					1	1	38	39	25,47
			2		1	3	176	177	18,03
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6 VI						0	178	179	20,11
						0	92	93	24,07
					3	3	162	163	24,21
13 IX					1	1	133	134	24,24
			1			1	233	234	24,10
	2					2	54	55	24,25
14 IX	4	1			1	6	76	77	24,13
					4	4	154	155	24,99
					0	0	68	69	24,93
16 IX					1	1	59	71	23,54
					2	2	226	227	22,88
17 IX						0	73	74	23,88
			2		3	5	106	107	24,22
						0	11	12	22,21
26 IX			1			1	169	184	23,65
Sum	6	1	7	0	21	35	$\left \right\rangle$	$\left \right\rangle$	495,02

Table 2 Summary	of flights co	nducted during	a tha summe	or of 2022
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The only instance of a recorded red deer is presented in Figure 1. The results from the summer UAV flights are considered less effective compared to those from previous seasons [9].



Figure 2 – Red deer recorded with a thermal imaging camera on September 14, 2022, during the second mission completed that day.

Examples of unrecognizable signatures are shown in Figure 3. The sample photos are from September 14 (a) and September 26 (b).



Figure 3 – Heat signatures detected that were not assigned to a species (a - 14 September, b - 22 September)

4. Discussion. Why the effectiveness of summer flights is so low?

According to Figures 4 and 5, it can be observed that most of the inventoried area was located in forest habitat types classified as mixed broadleaved (LM) and broadleaved (L) forests, covering approximately 76%. Together with the forest habitat type Mesic mixed coniferous (BMśw), where undergrowth is also often present, this accounted for around 97% of the total area. The dense and expansive tree canopies, together with the presence of understorey vegetation, account for the invisibility of animal heat signatures in the images taken by the UAV. It may be the case that in less fertile areas (coniferous habitat types), summer flights could be as effective as those conducted later in the autumn (after leaf fall) or in winter.

The forest habitat classified as "not specified" (Figure 5) includes elements such as fragments of forest roads, open areas directly adjacent to forests, small patches of private forest, or forest fragments for which there is no habitat types layer on the digital map used for spatial analysis (minor inconsistencies in the boundary lines of vector layers representing forest divisions and habitat types). We emphasize that the mentioned errors are negligible.

		·		SOIL FERTIL					
			Fertility (trophic) groups of forest habitats						
]	of forest habitats	CONIFEROUS (B)	MIXED CONIFEROUS (BM)	MIXED BROADLEAVED (LM)	BROADLEAVED (L)			
	DIENT	DRY	Dry coniferous Bs	-	-	-			
	RE GRA	MESIC	Mesic coniferous Bśw	Mesic mixed coniferous BMśw	Mesic mixed broadleaved LMśw	Mesic broadleaved Lśw			
	DISTU	MOIST	Moist coniferous Bw	Moist mixed coniferous BMw	Moist mixed broadleaved LMw	Moist broadleaved Lw			
Į	SOIL MC	SWAMP	Swamp coniferous Bb	Swamp mixed coniferous BMb	Swamp mixed broadleaved LMb	Alder OI			
		FLOODPLAIN	-	-	-	Riparian Lł \ Alder-ash OlJ			

Figure 4 - Percentage of forest habitat types over which a drone inventory was carried out (range from 0,03 % to 43 %)

The analysis of forest habitat types over which UAV flights are conducted is particularly important during the growing season, as more fertile habitats often have a more complex vertical forest structure. In single-storey stands within fertile habitats, the ground may also be covered by dense, expansive tree crowns (e.g. hornbeam). The thermal radiation detected in infrared images is suppressed not only by the upper canopy layer but also denser thickets than those found in poorer habitats. However, in this study, no detailed analyses were carried out on the impact of tree crown coverage on the results of game detection. In future research, this is undoubtedly a factor that should

be included in the analysis, as it is directly linked to the probability of detecting animals. Many other parameters should be considered, such as the height and age of the growing vegetation. The matter is further complicated by the fact that in the conditions of larger forest complexes, we are not able to determine the exact number of animals. There is no precise method, even a laborious one, that would allow obtaining accurate, reliable reference data.



Figure 5 - Percentage of forest habitat types over which drone surveys were conducted

In addition, it should be emphasized that the equipment used is commercially available. With a higher-quality thermal imaging camera, the results would certainly be much more satisfactory (e.g. clearer thermal signatures). The hardware solution used here should rather be considered affordable.

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References

- [1] Berni JA, Zarco-Tejada PJ, Suárez Barranco MD, Suarez L, Fereres E. Thermal and Narrowband Multispectral Remote Sensing for Vegetation Monitoring From an Unmanned Aerial Vehicle. *IEEE Transactions on Geoscience and Remote Sensing*, Vol. 47, No. 3, pp. 722-738, 2009, DOI:10.1109/TGRS.2008.2010457.
- [2] Bushaw JD, Ringelman KM, Rohwer F. Applications of Unmanned Aerial Vehicles to Survey Mesocarnivores. *Drones*, Vol. 3, No. 1, 2019, DOI:10.3390/drones3010028.
- [3] Choiński M, Rogowski M, Tynecki P, Kuijper DPJ, Churski M, Bubnicki JW. A first step towards automated species recognition from camera trap images of mammals using AI in a European Temperate Forest. *International Conference on Computer Information Systems and Industrial Management: Springer*, pp. 299-310, 2021, DOI: 10.1007/978-3-030-84340-3_24.
- [4] Chabot D, Bird DM. Wildlife research and management methods in the 21st century: Where do unmanned aircraft fit in?. *Journal of Unmanned Vehicle Systems*, Vol. 3, No. 4, 2015, DOI:10.1139/juvs-2015-0021.
- [5] Cilulko J, Janiszewski P, Bogdaszewski M et al. Infrared thermal imaging in studies of wild animals, *European Journal of Wildlife Research*, Vol. 59, pp. 17-23, 2012, DOI:10.1007/s10344-012-0688-1.
- [6] Corcoran E, Winsen M, Sudholz A, Hamilton G. Automated detection of wildlife using drones: Synthesis, opportunities and constraints. *Methods in Ecology and Evolution*, Vol. 12, pp. 1103–1114, 2021, https://doi.org/10.1111/2041-210X.13581
- [7] Engeman RM, Alien, L. Overview of a passive tracking index for monitoring wild canids and associated species. *Integrated Pest Management Reviews*, Vol. 5, pp. 197-203, 2000. DOI: 10.1023/A:1011380314051
- [8] Engeman RM, Witmer GW. IPM strategies: indexing difficult to monitor populations of pest species. Proceedings of the 19th Vertebrate Pest Conference, pp. 183-189, 2000, https://doi.org/10.5070/V419110013
- [9] Frackowiak R, Goraj Z. Animal detection using thermal imaging and a UAV. *Aircraft Engineering and Aerospace Technology*, Vol. 95, 2023, DOI:10.1108/AEAT-10-2022-0271
- [10] Witmer GW. Wildlife population monitoring: some practical considerations. *Wildlife Research,* Vol. 32, pp. 259-263, 2005
- [11] Gompper ME, Kays RW, Ray JC. et al. A comparison of noninvasive techniques to survey carnivore communities in northeastern North America. *Wildlife Society Bulletin*, Vol. 34, No. 4, pp. 1142-1151, 2006, DOI: 10.2193/0091-7648(2006)34[1142:ACONTT]2.0.CO;2.
- [12]Gooday OJ, Key N, Goldstien S. et al. An assessment of thermal-image acquisition with an unmanned aerial vehicle (UAV) for direct counts of coastal marine mammals ashore. *Journal of Unmanned Vehicle Systems*, Vol. 6, No. 2, pp. 100-108, 2018, DOI:10.1139/juvs-2016-0029.
- [13] Hamel S, Killengreen ST, Henden JA. et al. Towards good practice guidance in using camera-traps in ecology: Influence of sampling design on validity of ecological inferences. *Methods Ecology and Evolution*, Vol. 4, No. 2, pp. 105-113, 2013, DOI: 10.1111/j.2041-210x.2012.00262.x.
- [14]Jordan MJ, Barrett RH, Purcell KL. Camera trapping estimates of density and survival of fishers Martes Pennanti. *Wildlife Biology*, Vol. 17, No. 3, pp. 266–276, 2011, DOI: 10.2981/09-091.
- [15] Lahoz-Monfort JJ, Magrath MJL. A Comprehensive Overview of Technologies for Species and Habitat

Monitoring and Conservation. *Bioscience, Vol. 71*, pp. 1038-1062, 2021, doi:10.1093/biosci/biab073.

- [16] Lee, S. Song Y. Kil S.-H. Feasibility Analyses of Real-Time Detection of Wildlife Using UAV-Derived Thermal and RGB Images. *Remote Sensing*, Vol.13, 2021, https://doi.org/10.3390/rs13112169
- [17] Mou C, Zhu C, Liu T, Cui X. A novel efficient wildlife detecting method with lightweight deployment on UAVs based on YOLOv7. *IET Image Processing*, Vol. 18, pp. 1296–1314, 2024, DOI: 10.1049/ipr2.13027
- [18] Mullero-Pázmány M, Barasona JÁ, Acevedo P. et al. Unmanned Aircraft Systems complement biologging in spatial ecology studies. Ecology and Evolution, Vol. 5, No. 21, pp. 4808-4818, 2015, DOI: 10.1002/ece3.1744.
- [19] Pagacz S, Witczuk J. Estimating ground surface visibility on thermal images from drone wildlife surveys in forests. *Ecological Informatics*, Vol. 78, 2023, https://doi.org/10.1016/j.ecoinf.2023.102379
- [20] Pagacz S, Witczuk J. Wykorzystanie samolotów bezzałogowych i termowizji do nocnej inwentaryzacji kopytnych. *Studia i materiały CEPL w Rogowie*, Vol. 18., No. 4,pp. 50-57, 2016
- [21] Research Topic Report "Weryfikacja innowacyjnej metody inwentaryzacji zwierzyny grubej z wykorzystaniem bezzałogowych statków powietrznych (BSP) oraz narzędzi teledetekcyjnych, w powiązaniu z szacowaniem szkód łowieckich w wielkoobszarowych uprawach rolnych", Taxus SI Sp. Z o.o.- consortium leader, Museum and Institute of Zoology PAS, The University of Warsaw, Topic number 11/17, Online access: https://tbr.lasy.gov.pl/apex/f?p=102:3:::::P3_TEMAT:4231 (access from 14.10.2024 r.)
- [22] Sardà-Palomera F, Bota G, Vinolo C. et al. Fine-scale bird monitoring from light unmanned aircraft systems, Ibis, Vol. 154, pp. 177-183, 2012, DOI:10.1111/j.1474-919X.2011.01177.x.
- [23]Thompson J, Fleming P. Evaluation of the efficacy of 1080 poisoning of red foxes using visitation to nontoxic baits as anindex of fox abundance. *Wildlife Research*, Vol. 21, pp. 27–39, 1994, DOI: 10.1071/WR9940027