

# birdnetTools: An R package for working with BirdNET output

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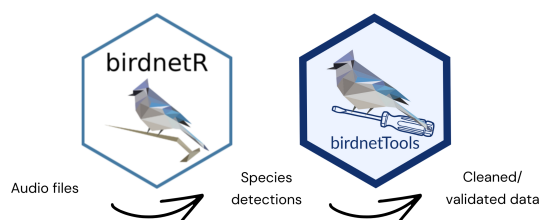
## Summary

birdnetTools is an R package for post-processing outputs from BirdNET, an open-source neural network developed by the Cornell Lab of Ornithology and Chemnitz University of Technology for detecting and identifying bird species from audio recordings (Kahl et al., 2021). The birdnetTools package streamlines workflows for cleaning and combining multiple BirdNET selection tables, filtering detections by species, confidence, or date/time, visualizing temporal and spatial patterns, and validating results using an interactive Shiny app. It also supports species-specific and universal confidence thresholds, enabling reproducible threshold-setting workflows.

## Statement of need

Automated acoustic monitoring is increasingly used in ecology and conservation (Pérez-Granados, 2023), with BirdNET being one of the most widely adopted tools for bird sound identification (e.g., Funosas et al. (2024), McGinn et al. (2023), and Bota et al. (2023)). Although BirdNET was developed in Python, most of its primary users are ecologists who conduct analyses primarily in R. This language difference can limit accessibility for some research teams. While the birdnetR package (Günther & BirdNET Team, 2025) enables R users to run BirdNET classifications, there is no dedicated framework in R for post-processing these outputs.

The birdnetTools R package fills this gap by providing functions built on workflows commonly used in published studies (e.g., Tseng et al. (2024)) and by incorporating threshold-setting and validation methods developed in recent research (i.e., Tseng et al. (2025); Wood & Kahl (2024)). By consolidating these tools, birdnetTools streamlines analysis and lowers barriers for ecologists and conservation practitioners adopting BirdNET in large-scale monitoring projects (Figure 1).



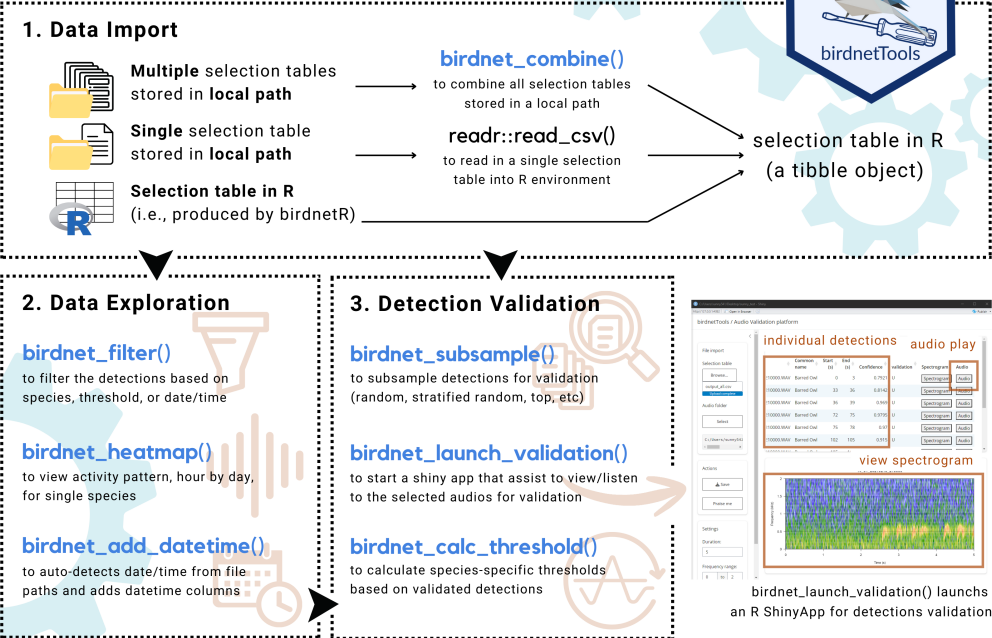
**Figure 1:** Integration of BirdNET R packages. birdnetR produces species detections from audio recordings, while birdnetTools provides tools for post-processing, data cleaning, and validation.

32 **Key functionalities**

33 Functions in `birdnetTools` fall into three categories: data import, data exploration, and  
34 detection validation (Figure 2).

- 35 1. Data import: `birdnet_combine()` integrates BirdNET outputs into R, supporting formats  
36 from the BirdNET GUI, Raven Pro, and the `birdnetR` package.
- 37 2. Data exploration: `birdnet_filter()` enables filtering by species, threshold, or date/time;  
38 `birdnet_add_time()` extracts temporal metadata; and `birdnet_heatmap()` visualizes  
39 activity patterns.
- 40 3. Detection validation: an R ShinyApp was developed, implementing threshold-setting  
41 approaches, including the precision-based method of Tseng et al. (2025) and the  
42 probability-based method of Wood & Kahl (2024).

**birdnetTools::work flow**



birdnetTools workflow updated 2025 Sep.15<sup>th</sup>

Figure 2: Workflow of the birdnetTools R package.

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## References

- 53
- 54 Bota, G., Manzano-Rubio, R., Catalán, L., Gómez-Catasús, J., & Pérez-Granados, C. (2023).  
55 Hearing to the unseen: AudioMoth and BirdNET as a cheap and easy method for monitoring  
56 cryptic bird species. *Sensors*, 23(16), 7176. <https://doi.org/10.3390/s23167176>
- 57 Funosas, D., Barbaro, L., Schillé, L., Elger, A., Castagneyrol, B., & Cauchoux, M. (2024).  
58 Assessing the potential of BirdNET to infer european bird species from acoustic data.  
59 *Ecological Informatics*, 164(112146). <https://doi.org/10.1016/j.ecolind.2024.112146>
- 60 Günther, F., & BirdNET Team. (2025). birdnetR: Deep learning for automated (bird) sound  
61 identification. In *Github repository*. GitHub. <https://birdnet-team.github.io/birdnetR/>
- 62 Kahl, S., Wood, C. M., Eibl, M., & Klinck, H. (2021). BirdNET: A deep learning solution for  
63 avian diversity monitoring. *Ecological Informatics*, 61. [https://doi.org/10.1016/j.ecoinf.](https://doi.org/10.1016/j.ecoinf.2021.101236)  
64 [2021.101236](https://doi.org/10.1016/j.ecoinf.2021.101236)
- 65 McGinn, K., Kahl, S., Peery, M. Z., Klinck, H., & Wood, C. M. (2023). Feature embed-  
66 dings from the BirdNET algorithm provide within-species classification of acoustic events.  
67 *Ecological Informatics*, 74(101995). <https://doi.org/10.1016/j.ecoinf.2023.101995>
- 68 Pérez-Granados, C. (2023). BirdNET: Applications, performance, pitfalls and future opportu-  
69 nities. *Ibis*, 165(3), 1068–1075. <https://doi.org/10.1111/ibi.13193>
- 70 Tseng, S., Hodder, D. P., & Otter, K. (2024). Using autonomous recording units for vocal  
71 individuality: Insights from barred owl identification. *Avian Conservation and Ecology*, 19.  
72 <https://doi.org/10.5751/ACE-02680-190123>
- 73 Tseng, S., Hodder, D. P., & Otter, K. A. (2025). Setting BirdNET confidence thresholds:  
74 Species-specific vs. Universal approaches. *Journal of Ornithology*, 166. [https://doi.org/10.](https://doi.org/10.1007/s10336-025-02260-w)  
75 [1007/s10336-025-02260-w](https://doi.org/10.1007/s10336-025-02260-w)
- 76 Wood, C. M., & Kahl, S. (2024). Guidelines for appropriate use of BirdNET scores and other de-  
77 tector outputs. *Journal of Ornithology*, 165. <https://doi.org/10.1007/s10336-024-02144-5>