

The WildlifeReID-10k dataset was created by using the Python package wildlife-datasets [25] by combining the following datasets: AAUZebraFish [7], AerialCattle2017 [5], AmvrakikosTurtles [2], ATRW [18], BelugaID [28], BirdIndividualID [10], CatIndividualImages [19], Chicks4FreeID [16], CTai [11], CZoo [11], CowDataset [12], Cows2021 [13], DogFaceNet [21], FriesianCattle2015 [6], FriesianCattle2017 [5], GiraffeZebraID [23], Giraffes [20], HyenaID2022 [29], IPanda50 [27], LeopardID2022 [29], MPDD [14], MultiCamCows2024 [30], NDD20 [24], NyalaData [9], OpenCows2020 [4], PolarBearVidID [31], PrimFace [1], ReunionTurtles [2], SealID [22], SeaStarReID2023 [26], SeaTurtleID2022 [3], SMALST [32], SouthernProvinceTurtles [2], StripeSpotter [17], WhaleSharkID [15], ZakynthosTurtles [2], ZindiTurtleRecall [8].

## References

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