

Visual Analytics of the Impacts of Climate Change on Migratory Bird Habitats Technical Document

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1 Introduction

Our Senior Capstone project is comprised of two components:

1. A species distribution model (SDM) capable of predicting how climate change will affect future bird habitats up to year 2100
2. A web-app that visualizes SDM output and displays other relevant information such as bird migration patterns and climate trends.

This technical document will overview how each of these components work, as well as the required data and file structure for them to operate correctly.

2 Species Distribution Model

2.1 Datasets

The species distribution model (SDM) uses two datasets, one for climate data and one for species distribution data.

Climate Data

Climate scenarios used were from the NEX-GDDP-CMIP6 dataset, prepared by the Climate Analytics Group and NASA Ames Research Center using the NASA Earth Exchange and distributed by the NASA Center for Climate Simulation (NCCS).

Thrasher, B., Wang, W., Michaelis, A. et al. NASA Global Daily Downscaled Projections, CMIP6. Sci Data 9, 262 (2022). <https://doi.org/10.1038/s41597-022-01393-4>

Thrasher, B., Wang, W., Michaelis, A. Nemani, R. (2021). NEX-GDDP-CMIP6. NASA Center for Climate Simulation. <https://doi.org/10.7917/OFSG3345>

Species Distribution Data

The occurrence datasets for our bird species come from the Global Biodiversity Information Facility (GBIF). The citations for each dataset is below.

****GBIF CITATIONS HERE****

2.2 R Processing

Libraries

- *dismo*: Methods for species distribution modeling, that is, predicting the environmental similarity of any site to that of the locations of known occurrences of a species.

- *sf*: Support for simple features, a standardized way to encode spatial vector data. Binds to 'GDAL' for reading and writing data, to 'GEOS' for geometrical operations, and to 'PROJ' for projection conversions and datum transformations. Uses by default the 's2' package for spherical geometry operations on ellipsoidal (long/lat) coordinates.
- *raster*: Reading, writing, manipulating, analyzing and modeling of spatial data.
- *sp*: Classes and methods for spatial data; the classes document where the spatial location information resides, for 2D or 3D data. Utility functions are provided, e.g. for plotting data as maps, spatial selection, as well as methods for retrieving coordinates, for subsetting, print, summary, etc.
- *ncdf4*: Provides a high-level R interface to data files written using Unidata's netCDF library (version 4 or earlier), which are binary data files that are portable across platforms and include metadata information in addition to the data sets.

Process

TODO

2.3 Model Training

Libraries

TODO

Input

TODO

Output

TODO

Process

TODO

2.4 Model Prediction

Libraries

TODO

Input

TODO

Output

TODO

Process

TODO

3 Web Application

3.1 APIs

React

Reactjs was used to create the frontend, and the main component is found under `src/App.js`. In `App.js`, there are calls to the backend to gather data for the various components located in the `src/components` folder. In the `components` folder, there is `BirdInfo.js` which returns the summary of the desired bird. There are four graphs: `ClimateChart.js`, `HeatMap.js`, `PolylineMap.js`, and `SDMchart.js`. Those four files contain the code to display the average temperature and prediction graph, the Trajectory graphs discussed more in the Leaflet subsection, and the png of the SDM output. Also in `components` folder is `PredictionControl.js` which contains the code to display the year slider and SSP buttons that change the Climate and SDM Charts. In addition to the components, the sidebar and header are created in `App.js` which allows users to view different birds. All the styling is written up in `src/App.css`.

FastAPI

The middleware and backend of the website was developed using python and FastAPI, in the file `base.py` located in the subfolder `backend/app`. The FastAPI backend primarily serves the purpose of quickly retrieving data and sending it to the front end as requested, allowing the front end user-interface to remain lightweight and easy to load quickly. FastAPI was used to build the backend due to its ease in creating end points for a frontend application to call, natively supporting features such as delayed requests and an intuitive way of passing arguments to the FastAPI backend. The backend is defined predominantly by functions that handle various requests the front end makes in order to retrieve data to display. These functions and their functions are discussed in greater detail in the following section for RestFUL API.

RestFUL

RestFUL API was used to define how the end points were set up, partially due to built-in compatibility with FastAPI, but also due to its simple and intuitive interface. While RestFUL API primarily defines 4 primary methods for GET, PUT, POST, and DELETE, the final application was operational with a stateless backend, meaning only GET and PUT requests were used. In the file `base.py`, GET and PUT requests were used to define the following end points:

- `get_temperature_data()`: a GET end point that in turn performs a POST request to `grid2.rcc-acis.org` to obtain temperature data for the state of California throughout a year, which is specified per the user's request.

- `get_precipitation_data()`: a GET end point that in turn performs a POST request to `grid2.rcc-acis.org` to obtain precipitation data for the state of California throughout a year, which is specified per the user's request.
- `get_predictions()`: a PUT end point that retrieves the pre-computed predictions the Species Distribution Model makes about a specified bird, for a specified year based on specified emission data. While a GET request would have sufficed for its current utility, a PUT request was preferred as this could be expanded to implement a machine learning model to generate predictions dynamically should such a feature be planned in the future.
- `get_trajectory_data()`: a GET end point that retrieves a csv file containing the migration trajectory of a specific bird, identified by its unique bird ID, that can be used to generate individual trajectory maps of an individual bird. The trajectory information is discussed later in the subsection for Leaflet.
- `get_bird_ids()`: a GET end point that returns all unique bird IDs per a given species. This is used to by the front end component `PolylineMap.js` to search for a specified bird ID.
- `get_heatmap_data()`: a GET end point that retrieves heatmap data of a bird species, to display the aggregated paths of all birds catalogued for that species on the front end with `HeatMap.js`, as a heat map. This is discussed in further detail in the Leaflet subsection.

Leaflet

Leaflet was used to map bird migration patterns in two ways under the "Trajectory" component. The first map, "Individual Path", is created in `src/components/PolylineMap.js`. The function fetches data from the csv files under the data folder in `backend/app`. The function first fetches all the tagged Bird IDs from the csv for the desired bird. Then, all the trajectories for the first Bird ID are fetched from the backend and drawn on `OpenStreetMap` using Leaflet. Arrows are calculated between two points in `calculateBearings` function and added onto the map as well. The second map, "Aggregated Path", is created in `src/component/HeatMap.js`. The function fetches all the latitude and longitude coordinates from the csv of the desired bird in the `backend/app/data` folder. Using Leaflet and the Leaflet plugin, `L.heatLayer`, a heat layer is generated showing the combined trajectory path of all the tagged birds on top of `OpenStreetMap` layer.

3.2 Usage

Insight into how the website should be used

3.3 Design

Insight into how the website was designed