WGFP_Enc_Summaries_RShinyApp How-To

Location:

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Table of Contents

Uses	3
General Notes	3
How to Open	4
Navigating the App	5
About/How to Use Tab	5
Individual Datasets Tab	5
Encounter Histories Tab	5
Encounter Release History Summary Wide	5
All Encounter Histories	8
Sequences	14
Examples	15
Weekly States Tab	21
States Dataframe	21
States and Days Wide	22
Unknown States	22
Daily Movements Map, Plots, and Data	22
"Movements" Defined	22
Map and Table	23
Plots	24
Pressure Transducer, USGS and Detection Distance	26
Time Series	26
Movements Overlay	27
Variable Correlations	27
Detection Distance	28
QAQC Tab	28
Marker Tags	28
Release and Recap Lengths/Weights	30
Unknown Tags	30
Ghost Tag Movements	30
Crosstalk QAQC	31
Detection Distance/Water Level	32

Adding/Removing Dummy rows in the data	32
Data Used in the App	33
Tabular data: WGFP_dataclean_vis2.0\data	33
Spatial Data: WGFP_dataclean_vis2.0\gis	36
Updating the App	36
File Readins	37
Updating file names	37
Other Data Updates	37
Creating New Flat Files/Running the Script	37
Common errors	38
Modifying, Updating, Adding New Antenna/Stations	39
Updating the code and data that wrangles the tabular data	39
Adding or updating layers on the map	43
Functions	45
All_Combined_events_function	45
Spatial_join_function	50
PrepareforStatesMovementsandSummary_function.R	50
Get_movements_function	56
Get_States_function	57
Shapefile/Polygon Readins	59
.Rds files	59
Advantages of .rds files in this app	60
Shiny Modules	60
Site, Movement and Species Colors	61

Uses

- Combines, cleans, and displays data from Stationary OregonRFID readers, Biomark antennas, mobile runs, Recaptures, release data, pressure transducers, site visits, and USGS in one location.
- Shows fish movement by day in map, table, plot, and custom animation
- Shows dataset of physical "states"
- QAQC of marker tags, lengths and weights, unknown tags, crosstalk, ghost tag movements

General Notes

- All plots are interactive. Hover over plot for more info and toggle what is displayed by clicking and double
 clicking items in the legend. Zoom in by clicking and dragging within the plot.
- Datatables are also interactive. Click the column titles of any column to sort in ascending/descending order. You can use the Search bar on the right to search/filter for any specific values. Most of the individual filters work, some don't, depending on column type I believe.
- It's important when adding a new updated file to the app that it had the exact same column names and order as the older file
- Data displayed in tables is often downloadable
 - When the table is accompanied by this download button, all data within the table will be downloaded



• When the table is accompanied by this download button, only the data that is displayed on the screen will be downloaded. Display all rows of the table to download all data in this table.

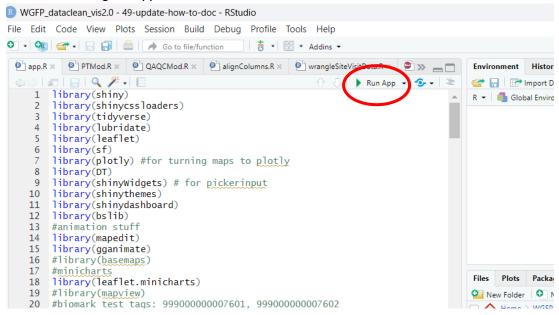


- UTM's for stationary and biomark stations are assigned in the metadata file
- DTY field for Stationary Antenna data refers to Detection Type in the raw .txt files but refers to date in the app

How to Open

OPENING FROM U: DRIVE

- I. Navigate to
 - U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections\CodingDetections\WGFP_Enc_Summaries_RShinyApp and open the WGFP_dataclean_vis "R Project" file. Rstudio will open
- 2. If the "app.R" file isn't already open in the top left panel: in the bottom right panel, navigate to the "files" tab and open the app.R file.
- 3. Click "Run App" in the top right of the screen. You may be required to download some packages if this is the first time running the app.



Note: It typically takes 20-40 seconds to read in all files from the app. If you are frequently opening and closing the app in one session, you can speed this process up by reading in the files to your local environment. These files are located in the "data read ins" section of the app.r file. To read them in, select all the code in this section, then use cntrl+enter to run this code. It takes 20-40 seconds to read in the files but once you do this, when you run the app each subsequent time, it will boot right up in a couple seconds.

```
♥ # Data Read Ins ----
  start_time <- Sys.time()</pre>
  print("Reading in Static Files for app.....")
if(!exists("combinedData_df_list")){
    combinedData_df_list <- readRDS("data/flatFilesforApp/combinedData_df_list.rds")</pre>
if(!exists("indiv_datasets_list")){
    indiv_datasets_list <- readRDS("data/flatFilesforApp/indiv_datasets_list.rds")</pre>
if(!exists("Enc_release_data")){
    Enc_release_data <- readRDS("data/flatFilesforApp/Enc_release_data.rds")</pre>
if(!exists("states_data_list")){
    states_data_list <- readRDS("data/flatFilesforApp/states_data_list.rds")</pre>
if(!exists("movements_list")){
    movements_list <- readRDS("data/flatFilesforApp/movements_list.rds")</pre>
if(!exists("Marker_tags")){
   Marker_tags <- readRDS("data/flatFilesforApp/Marker_tags.rds")</pre>
if(!exists("unknown_tags")){
    unknown_tags <- readRDS("data/flatFilesforApp/unknown_tags.rds")</pre>
```

Navigating the App

About/How to Use Tab

This tab is a concise version of this document. It doesn't have much there now but can be modified to include the most relevant pieces of the how-to doc.

Individual Datasets Tab

This Tab shows data incorporated into the app from different sources.

- Stationary Clean: A clean dataset of all Stationary antenna data with added UTMs, no marker tags, or duplicate
 rows. All timestamps and dates are in the same format. 900_ is taken off of the tags. See cleanStationary
 function in update data app for more info on cleaning.
- Biomark: All Biomark combined detections from Kaibab Park/Granby Diversion and both Windy Gap Sites including Marker Tags
- Mobile: All Detections from all Mobile Runs
- Recaptures: All Recaptured fish
- Release: All release data along with frequency plots for length and weight. Note: entries with no L/W listed won't show up in the plot
- Ghost Data: a list of ghost tags that also have a "ghost date" associated with them, or the date they turned into a ghost tag
- Aviation predation: list of tags with a "predation date" associated with them when they turned into a predated tag
- Pressure Transducers: a combined dataframe of all pressure transducer data in the app
- USGS 15 Minute Data: USGS Data from their website from the stream gauge at Hitching Post. As of 5/22/2024, only water temp and discharge are read in
- Site Visit Data: Combined site visit data from all stationary and biomark sites.

Encounter Histories Tab

Encounter Release History Summary Wide: shows ENC_Release_wide_summary Dataframe from enc hist summary wide function.

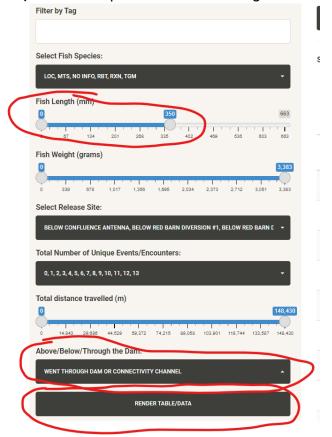
- This tab is meant to be a summary file for all fish in "wide" data format. Each tag will have just one entry, along
 with release info and information about what sort of Events it has experienced over its life.
- Filters are pretty self-explanatory. Click Render Table/Data to display new tables with filters.
- Should be same amount of entries as release file
- Columns and filters to note:
 - o SiteCode n is the raw number of detections at one antenna, or recaptured, etc.
 - SiteCode/Recapture binary columns are just whether or not a fish was detected or captured by that method
 - TotalEncounters adds the number of unique Events (a recapture, stationary detection, mobile detection, etc). There is a filter for this column in the sidebar.
 - Through_dam tells whether a fish has gone through the dam or through the connectivity channel in its history, based on release info, recaps, and detections. It does NOT tell if a fish went upstream or downstream through the dam. There is also a filter for this in the sidebar
 - CondensedAllStates is a quick history of the "states" dataframe (see Weekly States Tab) for each fish.
 Consecutive, identical letters are removed, leaving only the concise history. There is a filter for this in the sidebar.
 - o ChannelSummary: tells if a fish has had ANY detection on any CRCC channel
 - Went_above_dam_no_channel and subsequent columns are derived from the States function, which
 tells if a fish went below/above the dam using the channel or not from upstream or downstream,

compared to through_dam which simply says if a fish as gone through the connectivity channel or dam or not. These columns give more specific info than through_dam, telling how the fish got above/below the dam: whether through the connectivity channel or not, and whether they started upstream or downstream. For example, a "TRUE" value in went_above_dam_no_channel would mean that the fish started below the dam, made it above the dam, but didn't use the connectivity channel. Get-states function section for more details

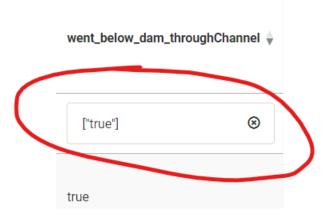
- Sum_dist is the total number of meters travelled by a fish in its history. There's a filter for this in the sidebar.
- This is a good tab to see how many fish have swam through CRCC, fish that have swam a long way, or find fish with extensive encounter histories while being able to filter on tag number, species, length/weight, Release Site.

Examples

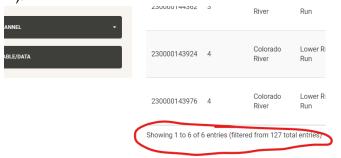
- Finding fish that have swam through the CRCC, specifically down through, that were less than 350mm.
 Strategy 1: Using Above/Below/Through_dam
 - a. Adjust the filters pictured to desired categories and render table



b. Scroll to the column went_below_dam_throughChannel and select "true" from the filters here



c. Scroll back to the beginning of the table and view the row numbers. There have been 6 fish of length 350 or below that have swam through the connectivity channel from upstream to downstream (as of 5/14/2024).



i. Note that the A fish registers TRUE in went_below_dam_throughChannel not solely if it is detected in the CRCC (see All Events and plot for that example), but only if the fish was actually detected in CRCC then subsequently detected (whether at RB, HP, Biomark, recapture, or mobile run) downstream of CD or CS.

Strategy 2: Using Condensed States

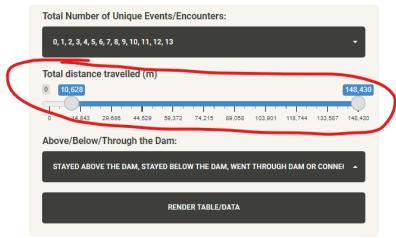
A fish will have the history "BCA" at some point (See "Weekly States tab") if they have been recorded above the CRCC (whether by mobile detection, recapture, biomark or stationary antenna), then detected by any of the CRCC antennas, then was recorded downstream of the channel (whether by mobile detection, recapture, biomark or stationary antenna). In the sidebar filter, find and select those instances, in addition to changing the "Fish length" filter



o Render this table, and here you can see again the 6 fish we found in strategy I



- This differs from finding "Sequences" in the CRCC (see "Sequences") because the states include recaptures, release, and mobile detections. However, you can't see timeframe with this data; use the Sequences tab or plug the tag into "ALL Encounter Histories" for that
- Identify Fish to explore as potentially avian predated
 - Fish that have swam abnormally long distances may have been avian predated. Adjust the distance moved filter and render table



Scroll to the "sum dist" column and click it to sort by ascending



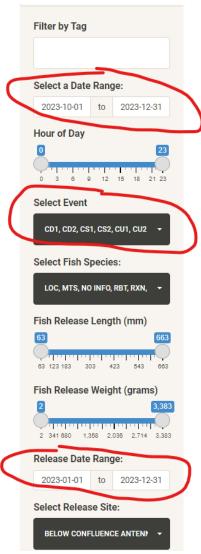
Scroll back to the beginning of the table to view the tag names, then manually view the encounter history of each tag in the movements tab or "All Encounter Histories" to see if their encounter history makes spatial and chronological sense. Lots of these tags may already be in the avian predation file, so check that file in the "individual datasets" tab to see if that tag has already been accounted for.

All Encounter Histories: Shows and plots All_Events Dataframe from All_combined_events_function. These are all detections/events combined, including Statoinary, Biomark, and mobile antennas; recaptures and releases. 230000142723 has over 120k detections on its own at Hitching Post.

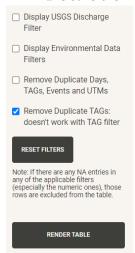
- Most filters are self-explanatory, but a the checkboxes are important to note:
 - Display USGS Discharge Filter allows you to filter data based off USGS discharge reading at Hitching Post within 13 hours of the detection (in winter, they record the data every 8-12 hours, 15 minutes in the summer). However, not every detection has an associated discharge reading because of this, so those rows get filtered out of the dataset immediately if you use this filter
 - O Display Environmental Data Filter allows you to filter data based off Pressure Transducer readings at specific sites within one hour of the detection. However, only sites that have a transducer at them will get this data associated with their detections and only within one hour, so using this filter will automatically filter out rows of data that do not have any of these readings associated with the detections.
 - Remove Duplicate Days, TAGs, Events and UTMs condenses detections into the first and last detections of the day as well as any other antenna detections with unique UTMs. In this way, it filters out most of the redundant detections across days. For example, using this filter condenses TAG 230000142723 down to 996 entries. This equates to about 259 days of detections from May 6, 2021 to Feb 3, 2022. All at Hitching Post.
 - The df that results of this filter used on all detections is downloaded, brought into GIS, and used to make stations with, which is then subsequently used to make the "movements" df with.
 - This is one of the most "useful" files you can get from this app because of how it shaves off the dataset to the most relevant detections.
 - o Remove Duplicate Tags Filter will remove instances where there are multiple detections of the same tag. This is helpful in answering questions involving "How many unique fish?". Example: on this tab you could answer the question "How many unique rainbow trout over 250 mm hit Windy Gap Biomark and Hitching Post antennas during the day in spring 2021?"
- The plot will plot anything made with the filtered data and the frequency table will display the raw number of detections from that filtered data for each antenna

Examples

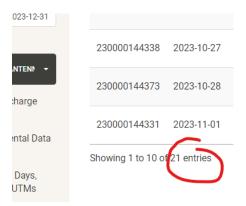
- How many fish released in 2023 were detected in the CRCC in Fall 2023?
 - o In the filters tab, change the following filters:



And make sure the "Remove Duplicate Tags" filter is checked before clicking "render table"



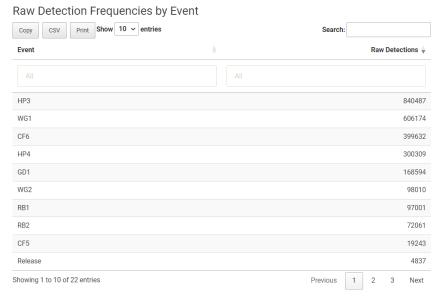
 The code filters the data based on the date range and events selected, then removes instances where the same tag shows up multiple times under these filters. The result is a df of unique tags that qualify for those filters. Find the number of fish by viewing the row number: 21



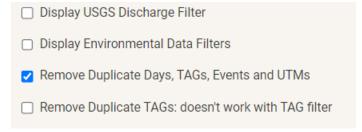
- Which antenna had the most raw detections in summer 2022?
 - Select a date range, render the table, and navigate to the "Plot and Table" tab



 Scroll to the "Raw Detection Frequencies by Event" table to see the distribution of events during that time frame: Looks like HP3 had the most raw detections.



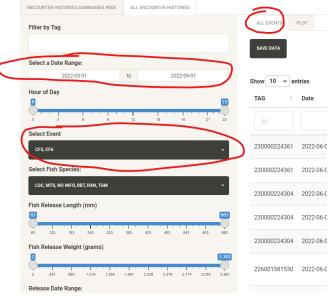
These numbers can shoot up if a fish is sitting on an antenna for a long amount of time. If we want to get a better picture of how much activity actually occurred on a daily level between antennas, we can use the "Remove duplicate days, tags, events and UTM's" filter described above.



The result shows a much better representation of what is going on at the sites across this timeframe, showing instances where fish had their first and last detections of the day and all detections in between. Hitching Post saw a lot of action.

Event		Raw Detections 🟺
All	All	
HP3		1398
HP4		1139
RB2		402
RB1		337
CF6		246
Release		218
GD1		132
Recapture		65
WG1		52
CF5		49
Recapture and Release		10
Showing 1 to 11 of 11 entries		Previous 1 Next

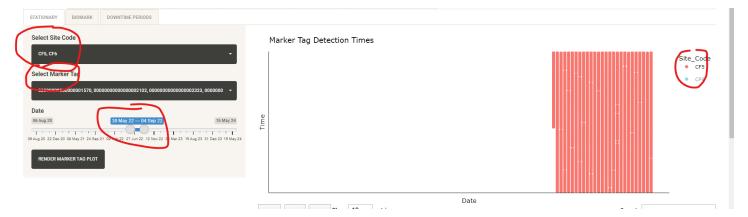
Why might CF5 have so much less action than CF6? To answer this, there's a few things to check. My first thought would be that there's a sculpin hanging out on just CF. We can go back to the "All Events" tab, select just those antennas in the "Event" filter, and scroll through the data so see if this is true



We see that there are a lot of different tags on CF6 during this timeframe, so that theory doesn't work.
 Next thought is if the detection distance was very low, or the antenna was just off during this time. To check that data, we go to the QAQC tab.



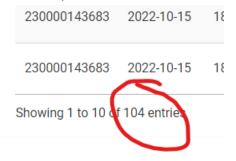
 Under "Marker Tags", can check if the antenna was running. Select CF5 and CF6 from the options, select all marker tags, change the timeframe to what we were looking at before, and render the data. Then toggle the plot displays by clicking in the legend.



- What we learn is that CF5 actually didn't seem to be working for much of this time period. If you were to turn on CF6 on the map, you'd see marker tag detections pretty consistently across the same timeframe. This explains the discrepancy between CF5 and CF6 during this timeframe.
- Seeing a Fish's Full Encounter History/Investigate Individual Tag for Avian Predation
 - Using one of the tags found (could be from by sorting on distance moved on the Encounter Histories Summaries Wide tab, seeing something in the states tab, seeing quick movement in the movements tab), we copy and paste that tag into the "All Encounter Histories" Tab. For example, I found 230000143683 (a big rainbow trout) by filtering for fish on Encounter Histories Summaries Wide tab that had I I or more "events".
 - Once the table is rendered, you can scroll to look at the fish's history (it default sorts by datetime). This fish has 630 encounters.



Scrolling through this data, you find yourself seeing a lot of detections on the same antenna across a bunch of days that doesn't tell you much. However, using the "Remove duplicate days, tags, events and UTM's" filter here is helpful again, preserving useful detections and removing ones that are unnecessary. In this case, it takes the total encounters down to 104.

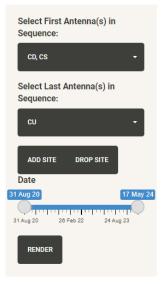


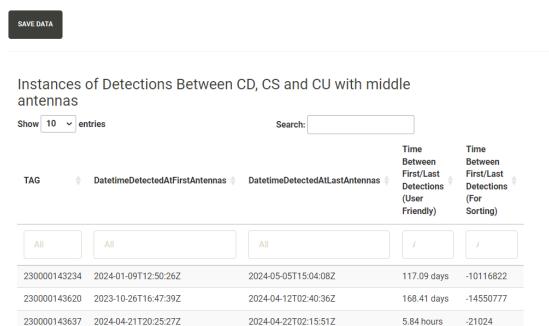
Now this is easier to view. The Encounter history of the fish started with its release 2022-10-5, hit WGI until the antenna was taken out of the river 2022-11-02. It hit hitching post and red barn next spring intermittently, including missing hp3 in late May, a time when the marker tag on HP3 wasn't being registered consistently (seen from the Marker tag QAQC tab). It was recaptured in June at lower Red Barn Fry Site and was registered as having grown 9mm in length. In December 2023, it made its way up the CRCC, hitting CD in December then resuming its journey in March, hitting CU March 12, then hit CF a week later where its more or less stayed since then (as of 5/14/2024). Based on this encounter history, I'd say the fish is not a tag succumbed to avian predation, just well-travelled.

Sequences

This is where you can pick a first antenna (or multiple), a last antenna (or multiple), and antennas in between then find instances where a fish have hit that sequence, and how long it has taken. This is an attempt to be able to easily answer questions like "How many times have fish used the connectivity channel?" or "are there any fish that have gone from Red Barn to the Granby Diversion Antenna?"

To use, select a downstream site (or more) and then select an upstream site (or more). Select Date range if desired. The table will then show instances where a fish has started at the first site(s) then was detected at the last site(s) with instances of middle detections in between. In the above example, this is helpful in showing instances where fish have moved through the CRCC (with CD, CS as first sites and CU as the last site).



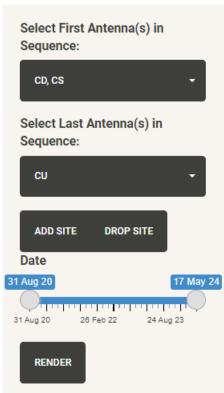


- fish that have traveled multiple times in the sequence will have multiple rows
- Ghost tags recorded in the ghost tags csv are included,
- I added a user-optional filter to include avian predation tags in results. Not all screenshots have this included because I added it later.
- Mobile detections, recaptures, and release data are NOT included
- A fish starts the sequence when it last hits the first site antenna and ends it when it first hits a site antenna from the last antenna
- For example, a fish could hit CD1 at 2023-11-03 12:04:15, hit CD2 on 2023-11-04 19:03:54, hit CS2 on 2023-11-06 03:34:41, hit CU1 on 2023-12-01 11:04:05, and CU2 on 2023-12-02 4:55:02. The Date/time sequence for this instance would be "2023-11-03 12:04:15" for the "DatetimeFirstDetectedatFirstAntennas" column, and 2023-12-01 11:04:05 for the "DatetimeFirstDetectedatLastAntennas" column. The time between US/DS detections would register as 28 days.
- When multiple antennas are selected for first and/or last, the sequence begins on the last instance of one of the first antennas. For example, if you have "CD, CS" selected for the Downstream antennas and a fish has hit CD2 on 2023-11-04 19:03:54, hit CU1 on 2023-12-01 11:04:05, and CU2 on 2023-12-02 4:55:02, the sequence will show up with 2023-11-04 19:03:54 in the "DatetimeFirstDetectedatFirstAntennas" column and 2023-12-01 11:04:05 in the DatetimeFirstDetectedatLastAntennas column.
- If there are other antennas that were hit in between the selected upstream sites and selected downstream sites, it breaks the sequence. So if you selected "HP" for first and "CF" for last, the results would not include fish that used the CRCC or hit WGI/2 along the way to get there. If you wanted to find these results, click "add site" and select all those antennas as the middle antennas.

- You cannot use a first/last antenna as one of the middle antennas, so sequences like "RB -> HP -> RB -> HP" aren't possible
- However, you CAN use the same antenna for first/last, so sequences like "RB -> HP -> RB" are possible to find.
- Use "All Encounter Histories" tab to plug in individual tags and investigate results you feel don't line up with what you'd expect to see in the results
- There's currently a bug that I'm not really able to replicate where sometimes you click "add site" and multiple inputs pop up off of that one click, instead of just one. If this happens, restart the app.

Examples

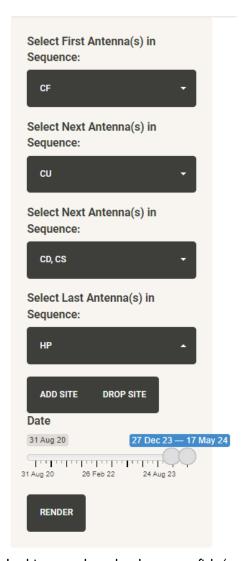
- How many time have fish have swam up the connectivity channel?
 - Selecting CD, CS as downstream antennas and CU as upstream antennas can tell you how many times fish have used the channel.



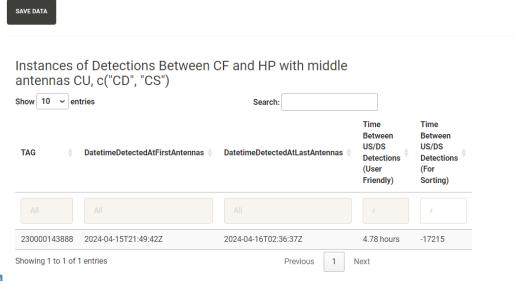
• Check the number of rows in the dataframe to get the answer: 20 times (as of May 27, 2024). A couple fish have made this journey multiple times.

Showing 1 to 10 of 20 entries

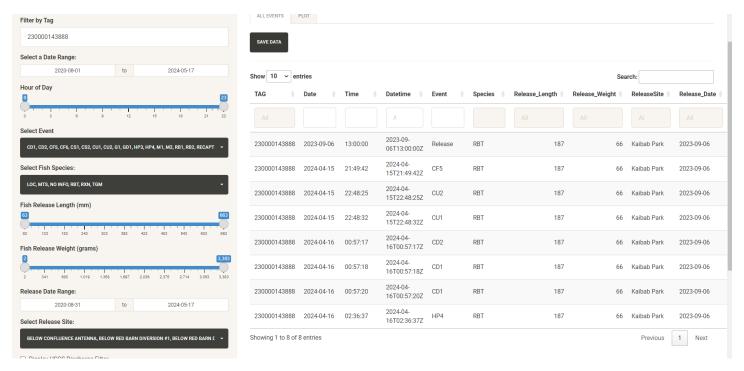
- How many times have fish swam from Confluence to Hitching Post through the channel in 2024?
 - Select the appropriate filters, using "Add Site" to include CRCC antennas in the desired sequence.
 - Note: There's a difference in selecting this sequence vs just selecting "CU, CD, CS" in just one middle antenna input. Selecting all the CRCC antennas in the same "middle antennas" input would allow for the fish to have a history like "2024-01-02 13:00:00 CF, 2024-01-03 15:00:00 CU, 2024-01-04 13:00:00 HP" and get included in the results dataframe when it misses CD/CS detections. The current inputs state that the fish have to first hit CF, then CU, then CD or CS before hitting Hitching Post. Another example: in a hypothetical scenario a fish could have this history: "2024-01-02 13:00:00 CF, 2024-01-03 15:00:00 CU, 2024-01-04 13:00:00 CS, 2024-01-08 13:00:00 CU, 2024-01-08 13:00:00 HP". With the current inputs, this instance would not be included in the results data, but if the inputs were instead changed to "CF", then "CS/CD/CU", then "HP", then this instance would be included since the fish just needs to have hit *any* of the selected middle antenna, not in a specific order.



o In this case, there has been one fish (as of May 27 2024).

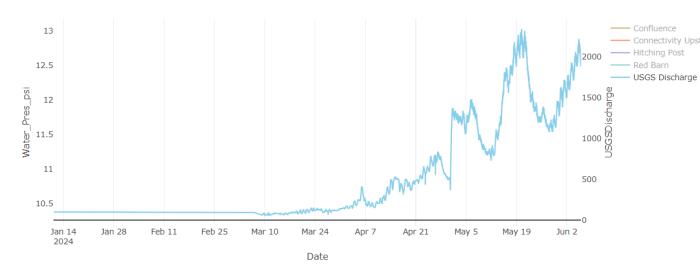


• We can plug this tag number into the "All Encounter Histories" tab or Movements tab to get more details about that instance.



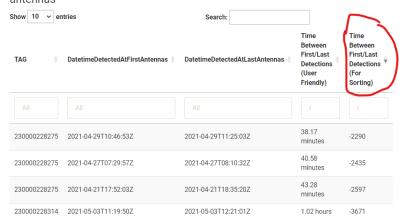
- One interesting takeaway is that the fish missed detection on CF6 and HP3 in its journey. A quick look at the marker tags data (see QAQC-Marker tags) shows us that the antennas were functioning well during that time. If you look at the time between the detections though, you'll see it took only 4.78 hours late at night for the fish to travel that far; it was moving super fast. Looking at the movements data –map and table tab in the "MPerSecond" column, we can see specifically that between CF and CU, the fish moved an average of 0.01 m/s, between CU and CD it moved .18 m/s, and between CD and HP it moved .34 m/s.
 - Note: on this fish since this movement spanned across multiple days, it has this more detailed speed travelled, but if the fish had just moved from CF to HP in one day (whether or not it hit CD, CU or CS in between), you would only see an average speed for that movement
- O A quick look at the discharge data (see "Pressure Transducer, USGS and Detection distance" tab) shows that flows measured at HP really started to rise around then, so it had some good momentum.

Environmental Time Series Data



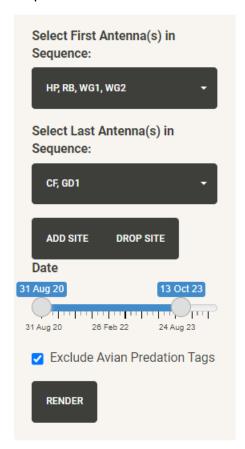
- What is the quickest time it's taken a fish to swim from RB to Hitching Post?
 - Select "RB" as the first antenna and "HP" as the last one. Render the data then click "Time Between First/Last Detections (for sorting)" to sort the results in either ascending or descending order.

Instances of Detections Between RB and HP with middle antennas

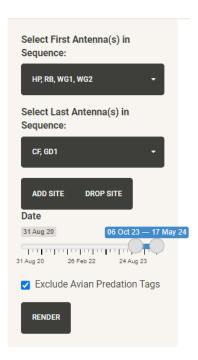


There was a fish that has gone multiple times in under an hour upstream from RB to HP. This is a bit of a red flag! Input the tag number into the "All Encounter Histories" tab to investigate further and record this in the avianpredation.csv if judged to be predation.

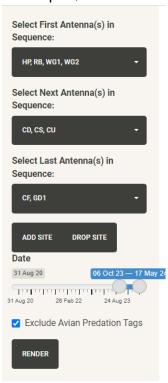
- How many fish have gone through the dam to go upstream since the connectivity channel was built vs before?
 - Set filters to below dam antennas for the first antennas, then the upstream of dam filters for the last antennas. Set the date range to "pre CRCC antenna construction" and filter out instances of avian predation



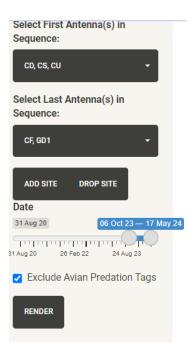
- Here we get 44 instances where non-avian predated fish* have gone upstream through the dam
 - *based on what is in the avian predation file. This tab could be used as another check to add tags to that file, especially using these filters
- Change the filters to post-CRCC construction



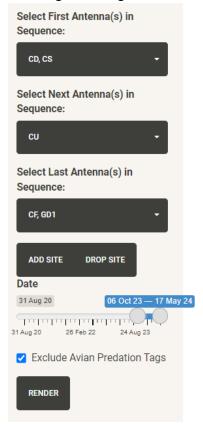
- Here we see only 3 instances of that sequence since CRCC was built
- o To compare, add a Site and adjust the filters as you wish



 In this instance, there's one instance of a tag recording this sequence. Since the CRCC hasn't been built long, a better measure might just be to exclude the below dam antennas.



 These filters yield 24 instances. However, it should be noted that this would include fish that might have swam down from CF, hit CU, then headed back to confluence. To get the more-desired sequence, the following filters might be most helpful.



- This yields 17 instances of fish who have used this sequence.
- o Interpret the data how you feel especially given how you can change the filters, but it's important to know what they are showing about the data. This would definitely suggest that fish are travelling upstream more now that the CRCC is built (17 instances vs 3 instances) but it's not exactly a straightforward comparison. When the Windy Gap Biomark antenna is consistently deployed, that may be a good way to level the comparison because then maybe we can see more clearly when fish come up to the dam, then decide which way of passage to use.

Weekly States Tab

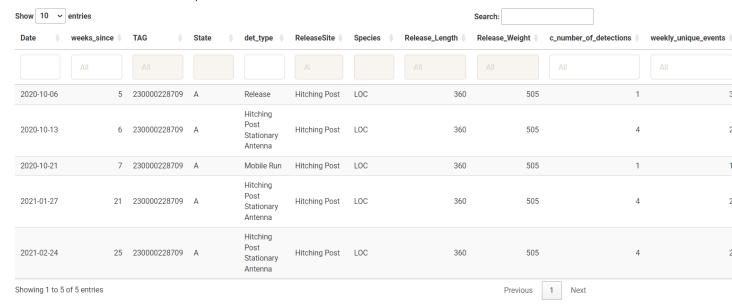
This is where "States" are displayed. The states are defined on a weekly basis as A (below the dam), B (above the Dam), C (in the Connectivity channel), G (ghost tag detection), or P (predated by bird). A weekly "Unique event" is a filter in the sidebar, and is defined as a detection on a new antenna for that week, recapture, or release. A fish could have 6 different events in a week, but if they are all below the dam, it will just get assigned state "A" for that week.

States Dataframe

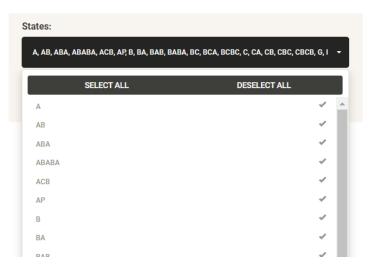
- The States filter in the sidebar can help determine "magic" fish that may have succumbed to avian predation, since they hit multiple weird states in a week. For example, 230000228444 has one week where it transitions at confluence, then red barn, then confluence again.
- Det_type column is useful when there is only I state of the week, but doesn't accurately capture everything if there are multiple states on a week
- Date is the day that the week starts, not when the actual detection(s) happened.
- C_number_of_detections is the total number of raw detections for the week. Might be 22k, might be 1.
- Weekly_unique_events is the total number of events that happened in a week. This can also be helpful in finding
 "magic" fish.
- These states are used to determine the columns "went_below_dam_throughChannel" etc in "All Encounter Histories Summary Wide" tab, because if a fish started upstream then used the channel, it will have B then a C somewhere in its history

Examples

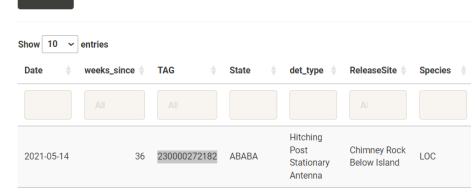
• Here is the fish 230000228709, a sizable brown trout.



- Finding Avian Predated Fish to Investigate
 - Use the "States" filter to view all the different states that fish have been in in a week. Select the most dubious like ABABA, BAB, etc.



Copy Tag number and investigate in the "All Encounters" or "Movements" tab



Showing 1 to 1 of 1 entries

States and Days Wide

This is the same data displayed in States_dataframe, but in wide format where each unique tag gets a row and each column is a day since the beginning of the study. This may have a bit more rows than the release file because there are some fish in there without release info, but should typically have the same

Unknown States

This is where events that were unable to be captured by states are displayed. This hopefully should be pretty small...it's filled with tags with detections before official 'Release' such as in in May 2021 and tags without release info. Mainly tags without an idea where they came from. But if a Tag shows up where release info is known, might have to go into the get_states_function.R code to make another case_when entry to account for the new state. All in all, this is a check to see how well the get_states_function is working, and also to see if fish without release info have gotten by the other filters.

Daily Movements Map, Plots, and Data

Displays data of "movements" (Movement_table_no_transitions from Get_movements_function), where a fish has changed position along the river.

"Movements" Defined

Movements are inferred by encounters of the same tag on different antennas (including mobile runs) or recaptures throughout the river. Each detection along the river is assigned a "station", breaking up the centerline of the stream into 10 m increments. The stations start at 0 m, starting about 150 m below the Sherriff Ranch Fry Site. They extend to 13480 m on the Upper Colorado River, where the mobile run begins, and to 22850 m (10120 m at confluence + 12730 m from confluence to Fraser River canyon) on the Fraser River (5/23/2024).

These stations, which are made in GIS from stream centerlines, are joined by coordinates to a dataset of all relevant detections for each unique UTM, using the Spatial_Join_function. The resulting dataset is passed back to a larger dataset of all detections to assign stations to all UTMs. From there, the data is condensed down to a dataset of daily fish detections, keeping the first and last detections of the day and one detection in between. Wacko fish/avian predation that hit multiple different antenna in between the first and last detection of the day won't display all of these detections. Example: tag 230000272182. These CAN be seen in the states and All Events dataframes though.

Once there is a dataset of daily movements, the distance moved between days is calculated for each fish when a new detection is recorded, calculated by taking the difference between stations.

Types of Movement

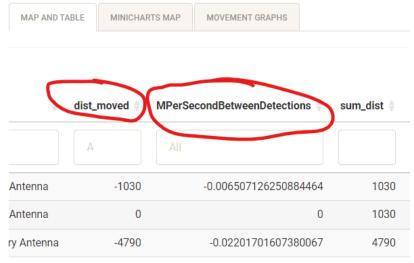
- Upstream movement: positive difference between stations (IE new detection station = 8000, previous detection station 6000, so 8000-6000 > 0; upstream movement).
- Downstream movement: negative difference between current station and previous station (new station = 6000 m, old station is 8000 m, -2000 <0; negative movement).
- No movement: the previous station and current station are the same (IE new station = 8000, previous station = 8000. Difference = 0, no movement). Consecutive detections at the same stationary antennas spanning more than one day are registered as no movement.
- Changed Rivers: Where a fish has moved from the Fraser River to the Upper Colorado (above the confluence) or vice versa. (IE a fish was detected on a mobile run on the upper Colorado but its release site was Fraser River Ranch. Although this is no upstream/downstream movement, the total distance moved in one example is 1910 meters, since that is the summed distances of Fraser River Ranch to the confluence + confluence to gps coordinates where the fish was detected on the Upper Colorado mobile run).
- Initial Release: Event when the fish was released. Used to establish the first station on the river that a fish was seen.

The absolute value of total distance moved is summed for each fish and displayed in the movements tab under "sum_dist". It also seen in the Encounter Histories created in the Ind_tag_enc_hist_summary_wide function.

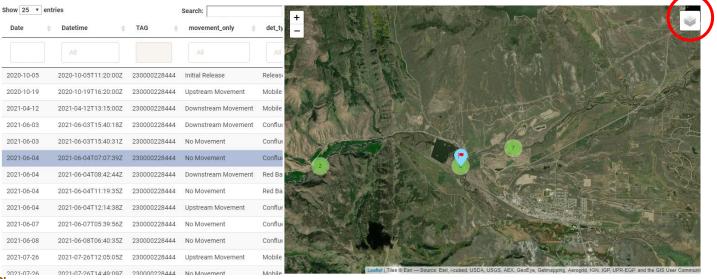
- Filters in the sidebar control what is displayed on the map or plot
 - USGS Data is combined with movements on a daily level, contrasted with events in All Encounter
 Histories where PTdata and USGS data is associated with an event within 13 hours for USGS data and 1
 hour for PTdata
 - Like the Environmental Filters and USGS data filters in Encounter Histories Tab, using the filters for Release length, USGS WaterTemp, and USGS Discharge will automatically filter out rows that don't have that data associated with the movements. All the Tiger Musky in the release file for example don't have release length/weight data so they don't show up on the map/table/plots if the release length filter is on

Map and Table

- This tab can be used to answer questions like "what time of year are fish moving?" and is the easiest/funnest way to examine individual tag histories. A method I frequently use is to find fish that have moved a long distance or have weird daily states recorded, copy and paste the tag, and plug it into this tab. Or find an event on the map, click on it, copy the tag number, and plug that in.
- Map is interactive, where you can click and hover to get more info about a movement. Clicking on an event in the map will navigate to the event column in the table, and vice versa.
- Can be used to help find suspect avian predation tags by clicking/sorting on the dist_moved and MetersPerSecondBetweenDetections columns

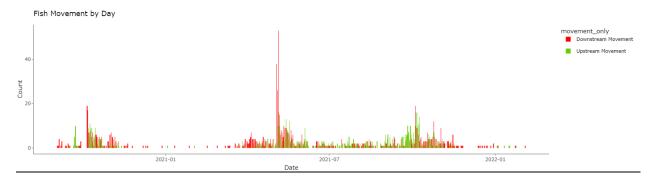


Layer control is available in top right, controlling detections, antennas, Release Sites, Stream centerlines, Stations
that were used to make these movements, and mobile reaches. Note that loading the Stations takes about 10
seconds and can also slow map performance.



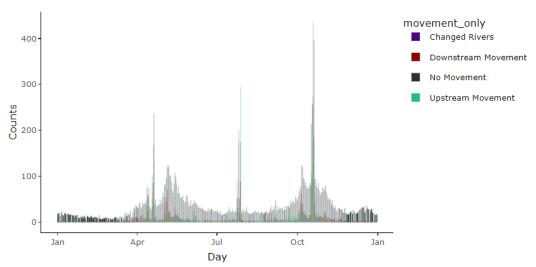
Plots

- The plots are also controlled by the sidebar filters
- See "types of movements" above for movement definitions
- Fish Movement by Day: simple histogram representation of movements across time, colored by movement type
 - Note: data can be misleading if mobile detections aren't filtered out, as these cause big spikes in movement activity the day of the mobile run. Activity is also generally higher after release events as well.

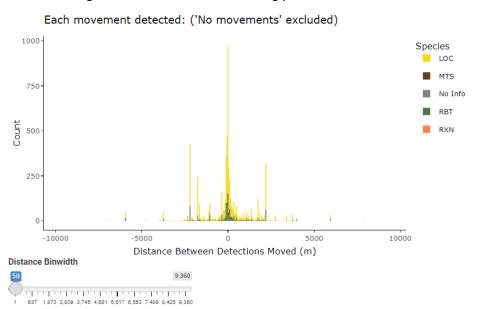


 Seasonal Daily Movements: Same plot as above but grouped by day across the years, which is helpful in seeing seasonal movement trends (big spikes here are probably mobile runs)

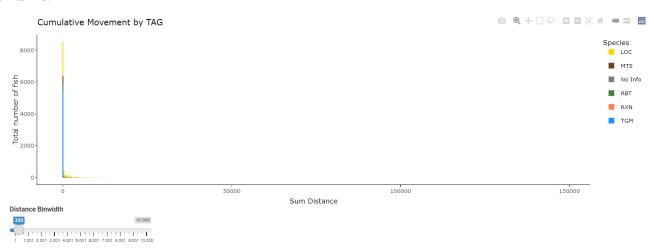
Seasonal Daily Movements



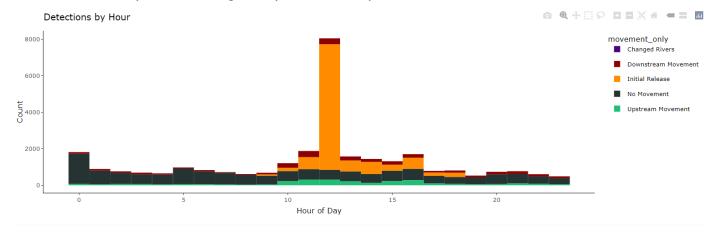
• Individual Movements: showing all the times a fish has moved (upstream, downstream or changed rivers) upstream and changed rivers are positive integers, downstream is negative. Note the spikes around 2200m due to fish travelling between red barn and hitching post, a distance of 2190m according to the stations.



• Cumulative Movement: the amount each fish has moved in absolute distance, in meters, in total over the course of its life.

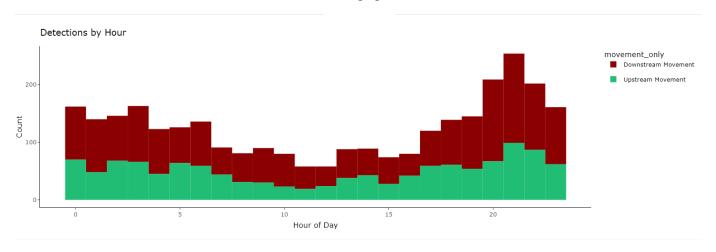


 Movement by hour: displays how many movements have been recorded during each hour of the day. Can be skewed towards hour 0 (12:00 AM) especially if a fish is hanging out on an antenna, since the movements df take the first and last activity of a fish during the day as well as unique events in between.



Data may be misleading; Movements are calculated by the day and if a fish has multiple detections on the same day at the same antenna, the first detection will be used which can fall commonly at hour 0

o If "No Movements" and "initial release" are filtered out from movement type, and "mobile detections", "release" and "recaptures" are filtered from movement types, you get movements registered by antennas in the river from fish that are not hanging out on antennas for weeks at a time.



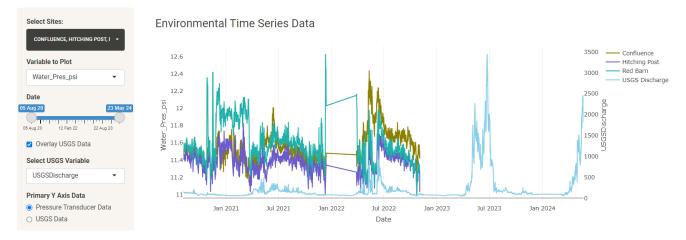
Pressure Transducer, USGS and Detection Distance

This tab shows graphical representation and overlays of data from Pressure Transducers, USGS and Detection Distance from Site Visits.

- Water_level_Nolce is the best variable for water level. Ice readings/observations are displayed as 0.
- USGS data is all from the gauge at Hitching Post. It can be misleading if you select a site and see discharge, but it's all from the same gauge.

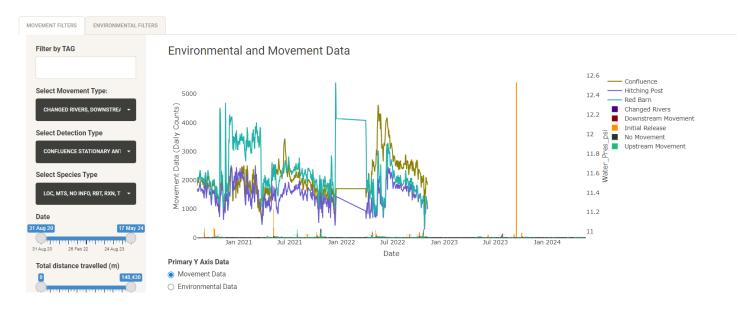
Time Series

PT data with the option to overlay USGS data. You can select variables to plot and which Y axis to display it on



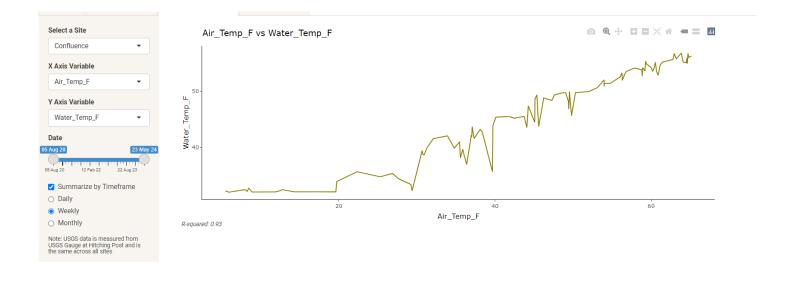
Movements Overlay

PT/USGS data overlaid with the same movements df displayed and used in the movements tab. All the same filters are available for movements, and the graph used is the Fish Movement by Day histogram of movements across time. Use the tabs in the sidebar to get to different data filters.



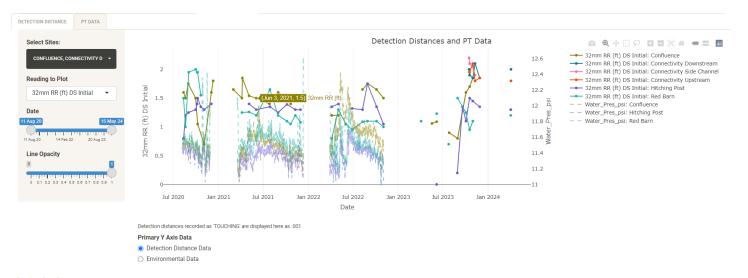
Variable Correlations

Plot PT/USGS variables against each other. "Summarize by timeframe" is an option used to group data by specified timeframe and can get rid of daily fluctuations. R-squared value in the left graph corner is calculated and found with this function in R: rquared = summary(lm(variableX ~ variableY))\$r.squared



Detection Distance

Detection distance from site visits of Biomark and Staionary Sites with PT data overlaid. You can select which reading you'd like to plot from the sidebar. You can change the opacity of the lines for preferred representation as well.



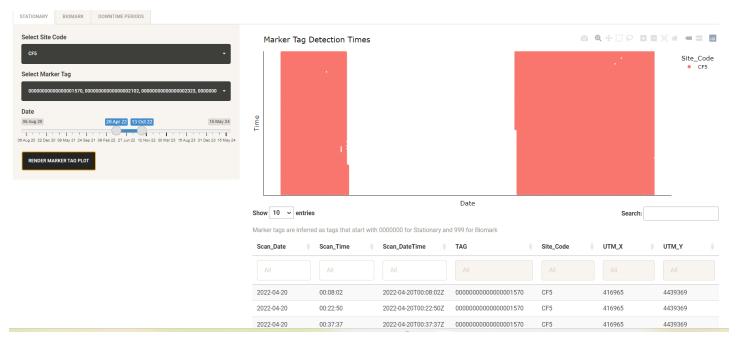
QAQC Tab

Used to help verify data.

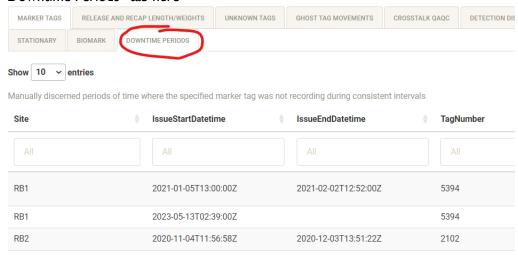
Marker Tags

Stationary and Biomark Marker Tag Data set from All_Combined_events_function

- Helpful for finding periods of inconsistent marker tag detections, indicating antenna was not functioning properly during these time.
- In this case, there was a gap: closer examination shows there were not detections between May 21 2022 and August 9 2022 at CF5



• These found gaps are manually recorded in the "Marker Tag Issues" tab in the metadata and are displayed in the "Downtime Periods" tab here



- This is a large dataset to plot with a scatterplot, so it will run quickest when plotting a smaller date range with only one marker tag or site selected.
 - A smaller date range will get better resolution in the plot, making it easier to see downtime periods with the tag.
 - What I do to find these periods is examine a small date range of about 2-8 months, select just one site, then select all marker tags and render the plot.
- If the ARR column is brought in incorrectly from stationary data (which can sometimes happen if the stationary ARR column gets corrupted, rarely happens) then there will be an abundance of detections in the 12:00-13:00 range.

Summarized Marker Tag data

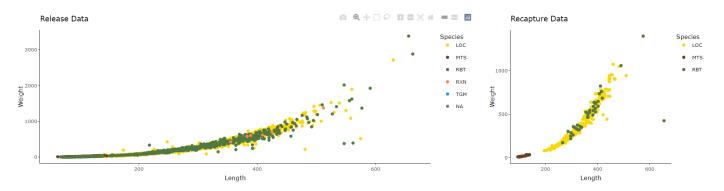
- Raw counts of marker tag detections across selected sites for the duration of the study.
- Helpful for seeing which tags have been deployed at which sites and which sites may have had more downtime periods than others

Summarized Marker tag Data for Selected Tags and Sites



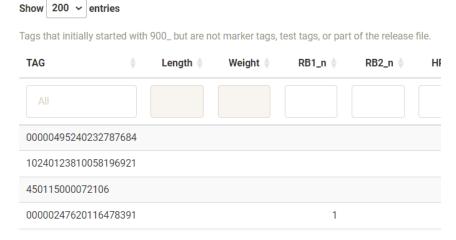
Release and Recap Lengths/Weights

• Plot to help ensure that there were no length/weight typos in the release and recapture files.



Unknown Tags

- List of Tags without release info but started with 900 initially and have detections on some sort of antenna
- Display of Enc hist wide summary function output "Unknown Tags"



Ghost Tag Movements

A df made from the "Ghost Tags" csv input. This shows registered ghost tags that have moved >0m after their assigned ghost date. This is helpful in revising the ghost tags file if one erroneously turns out to be a fish.

• Table is automatically sorted by "total_distmovedAfterGhostDate" so the most problematic-looking fish are first displayed

- antennasDetectedAfterGhostDate is a vector of all antennas that have detected the tag after the ghost date. If
 the tag was identified as a ghost tag somewhere near upstream of a stationary antenna, it isn't uncommon to see
 the tag be detected on that antenna during a high flow event
 - To get more context, you can check USGS discharge in the "Pressure Transducer, USGS, and Detection Distance" tab, find the discharge associated with that specific detection within 13 hours in "Encounter Histories All Encounter Histories, or find the daily discharge associated with each daily movement in "Daily Movements" tab Map and Table "USGSDischargeDaily" column
- If a fish is moving upstream significantly after its ghost date, it might be worth investigating. This is seen by sorting/clicking the "maxUpstreamDistMovedAfterGhost" column.
- Movements of 10-50m ish between mobile runs can typically be attributed to GPS variation

Ghost Tags To Double Check: each of these tags has moved > 0m since its assigned Ghost Date.

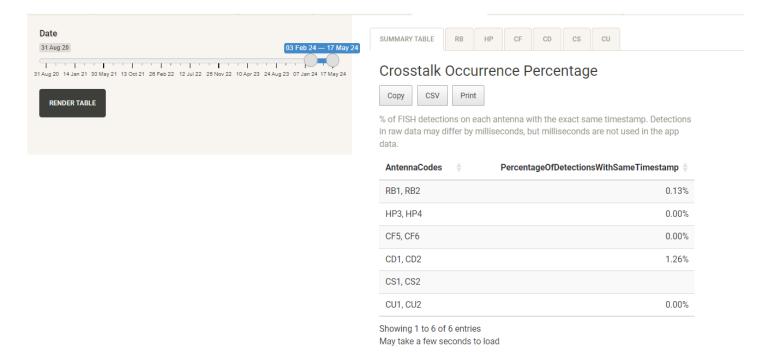
TagID	GhostDate	antennasDetectedAfterGhostDate	total_distmovedAfterGhostDate	maxUpstreamDistMovedAfterGhost	Notes
All	All	All	All	All	All
230000272949	2021-07-28	Mobile Run	230	0	Likely Ghost; Big fish location and probably date of 7-28-21); ghos 26-23
230000142602	2021-07-28	Mobile Run	220	20	confirmed using encc
		MITTER TRAIN B. COLUM			0 6 101 111

Crosstalk QAQC

Marker Tag crosstalk, where the tag goes off on one antenna but also registers on the other antenna as well, has been occurring frequently on CD since its installation. This tab is meant to monitor crosstalk instances of *Fish*, where a non-marker tag is registered at both antennas of a site at the exact same time.

Is created from combinedData_df_list\$All_Events, originally stemming from All_combined_events_function().

- Summary Table shows the percentage of fish detections across the selected timeframe have been recorded at the exact same time
- Individual site tabs show when those simultaneous detections occurred
- There are instances where the simultaneous detections occurs naturally. In the app, milliseconds are not used, but the data coming off the readers do use milliseconds. Sometimes there are instances where a fish is recorded milliseconds apart on different antennas, but these will show up as the exact same timestamp in the app. However, when CD is doing it far more frequently than the other sites, that probably means something more is going on.



Detection Distance/Water Level

Shows a color coded table of SiteVisitAndPTData, a joined df of PTdata and Site Visit Data by time to the nearest 13 hours.

- Using the variable "Water_Level_Nolce_ft", this table colors red the 32mm readings that are *below* the observed water level and green *above or at* the observed water level
- Helpful for identifying periods where detections may have been missed because of high flows.
- Readings that were entered as "TOUCHING" are changed here to .001



Notes: water level data is calibrated to staff gage at antenna site, but actual depths across antennas will vary

Adding/Removing Dummy rows in the data

Dummy rows were added to the stationary, biomark, and release data using the dummy_rows.R script to ensure the framework for new antennas was in place.

The rows remain while the function is ran to set the framework then the rows are removed in the data so this dummy data isn't used. The rows are removed....

In the get movements function

```
movement_table_notrans <- combined_events_stations %>%
  9
         #### removing dummy tag
         filter(!TAG %in% c("230000999999")) %>%
 10
         select(Date. Datetime. TAG. det type. Event. ET STATION. Si
In the reactive for all events in app.r:
  919
  920 -
  921
             #### filter dummy row
              all_events_filtered <- all_events_filtered %>%
  922
               filter(!TAG %in% c("230000999999"))
  923
  924
  925
  926
               return(all_events_filtered)
  927 -
          }) #end of ENC data list eventReactive
  928
                           #daily_unique_events = length(unique(Event))
                     30
                     31
                           states1 <- wghost_av %>%
                             filter(!TAG %in% c('230000999999')) %>%
                     32
In get states function:
                    33
                             mutate(
```

After the functions are ran in the following individual datasets in app.r:

```
### taking dummy tag out

152
153    Biomark <- Biomark %>%
154    filter(!DEC.Tag.ID %in% c("900.23000099999"))
155    Release_05 <- Release_05 %>%
156    filter(!TagID %in% c("230000999999"))
157    Stationary <- Stationary %>%
158    filter(!TAG %in% c("900_23000099999"))
159
160    # Define UT for application that draws a histogram In Ind_tag_enc_hist_wide_summary function:
```

dummy rows removal: 1/14/23

ENC_Release6 <- ENC_Release6 %>%

To put the dummy rows back in, comment out the code that filters out the dummy rows above.

filter(!TAG %in% c("230000999999"))

Data Used in the App

171

172

173 174

The following files are used in the app and are manually derived.

```
Tabular data: WGFP_dataclean_vis2.0\data
Stationary detections: WGFP_Raw_yyyymmdd.rds
```

- Combined and cleaned file of all stationary detections. Obtained from the "Combine data RShiny app" found at U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections\CodingDetections\WGFP_CombiningData_RShinyApp.
- Most recent file is found at
 U:\Projects\Colorado River\Windy Gap FishMovementStudy\Data\RFID\Detections\All Stationary

Mobile: WGFP Mobile Detect AllData.csv

- A combined file of all mobile detections found at U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections\MobileRaftAntenna\Mobile Detections
- Usually Eric R prepares this
- Column names and order: "Num", "River", "MobileSite", "Date", "Time", "T109_C", "UTM_X", "UTM_Y", "TagType", "TagID", "Event", "Ant", "Pass", "Species", "Length", "Weight", "TagSize", "RS_Num", "ReleaseSite", "Survey", "Notes"
- Of these, TagID, Date, Time, UTM_X, UTM_Y, and Ant are the most important columns

Biomark: Biomark Raw yyyymmdd.csv

- Combined file of all Biomark detections, found at
 U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections\Biomark
- Made from the "Combine New Data RShiny App"
- Column names and order: "Scan.Date", "Scan.Time", "Download.Date", "Download.Time", "Reader.ID", "Antenna.ID", "HEX.Tag.ID", "DEC.Tag.ID", "Temperature.C", "Signal.mV", "Is.Duplicate", "Latitude", "Longitude", "File.Name"
- Most important columns are Reader.ID, Scan.Date, Scan.Time, DEC.Tag.ID

Release: WGFP ReleaseData Master yyyymmdd.csv

- Master Release file of all tagged fish. Found at
 U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Tagging
- Column names and order: "RS_Num", "River", "ReleaseSite", "Date", "Time", "Year", "UTM_X", "UTM_Y", "Species", "Length", "Weight", "TagType", "TagID", "QAQC", "TagSize", "Ant", "Event", "FinClip", "Mortality", "Comments"

Recaptures: WGFP RecaptureData Master yyyymmdd.csv

- File of all fish that were recaptured found at
 U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Recaptures
- Column names and order: "RS_Num", "River", "RecaptureSite", "Date", "Time", "UTM_X", "UTM_Y", "Species", "Length", "Weight", "TagType", "TagID", "QAQC", "TagSize", "Ant", "Event", "FinClip", "Mortality", "Comments"

Avian Predation: WGFP AvianPredation.csv

- csv of tags succumbed to avian predation with a date of predation found at
 "U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections\AvianPredation\WGFP_A vianPredation.csv"
- used in the "States" function to assign a predated state

- "U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections\GhostTags\WGFP_Ghost Tags.xlsx" Need to export/save the "Ghost tags" tab to csv
- csv of ghost tags with date of ghost tag
- used in the "States" function to assign a ghost state
- Column names: RS_Num, River, ReleaseSite, ReleaseDate, Species, Length, Weight, TagID, TagSize, Event, GhostDate, UTM X, UTM Y, Comments

Metadata: WGFP Metadata.xlsx

- Found in U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections
- Used to keep track of antenna names, UTMs and what they're called in the data, as well as test tags, marker tag downtimes, and important stationing variables.
- Includes tabs:
 - O Antenna Metadata: Used to help assign names and UTMs to the antennas
 - AntennaSite: Name of Antenna (ie Red Barn (Upstream))
 - SiteName: Name of Site (ie Red Barn Stationary Antenna). DO NOT CHANGE THIS
 FIELD for an antenna
 - FrontendSiteCode: How we want the shorthand code to display in the app. Needs to be just one entry but can be changed as you wish.
 - BackendSiteCode: code(s) that have been used in the data to denote this antenna. Make sure this code is the same ones that come off the readers: SCD field for Stationary, Reader ID for Biomark. It's ok to have multiple names for the Biomark Antennas separated by comma and a space, but not for Stationary ones
 - PressureTransducerSiteName: Name of the Site as it appears in the Pressure Transducer Data.
 Must be updated if new sites are added
 - DetectionDistanceSiteName: Name of the site as it appears in Site Visit/Detection Distance
 Data
 - UTM X and UTM Y: UTMs of the site
 - River: River it's deployed on. DO NOT CHANGE THIS FIELD for an antenna
 - Deployment Duration: Dates and times where they have been deployed
 - Notes: any notes on the antenna
 - MarkerTagIssues: used to keep track of downtime periods on the antennas. Displayed in the QAQC tab of Marker Tags under "Downtimes"
 - ImportantStationingVariables: Used to keep track of important sites, used for movement calculations within the app
 - o TestTags: Keeps track of all test tags and removes these tag detections from the data
 - O Notes: Keeps track of general metadata in the worksheet. Current notes: (5/24/2024)
 - Assumes that for AntennaMetadata, SiteName and River will not be changing. If they do change, you'll have to go into the runscript and a couple functions including
 "PrepareforMovementsStatesand Summaries" to change how those variables are located
 - For the biomark antennas, I don't think you can actually code the readers to detect as the frontend codes; only codes like A4,B2 etc...hence the need for frontend/backend codes
 - For antenna metadatea codes, the frontend code should only have I entry. It's ok for the biomark backend codes to have multiple entries, but not the stationary ones
 - For marker color to be assigned correctly, need to have "Biomark Antenna", "Stationary Antenna", and "Mobile Run" in SiteName
 - First data date: 2020-08-06. detections before this date are removed
 - PressureTransducerSiteName needs to line up with the sites that are in the Pressuretransducer data

Pressure Transducer: WGFP_PressureTransducer_sitename.csv

U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\Pressure_Transducer

- Cleaned Pressure Transducer files from each site in csv form are all put into \WGFP dataclean vis2.0\data\PressureTransducer
- ALL FILES MUST HAVE SAME COLUMN NAMES/ORDER

Site Visits: WGFP SiteVisits FieldData.xlsx

- U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\SiteVisits
- 2 relevant sheets, Stationary and Biomark used for detection distances.

Spatial Data: WGFP dataclean vis2.0\gis

Antenna_sites I.shp: point file of antenna sites locations

Stream Centerline Post.shp: centerline of the Fraser, Upper Colorado, and Colorado River below windy gap

ReleaseSites2021.shp: point file of release site locations

mobile reaches.shp: mobile reaches extents

simpleStations.rds: stream centerlines broken into 10 meter sections. Originally from Stations_10m_Post.shp, but simplified to 10% of original resolution to improve map loading efficiency (see

Updating the App

The app data is updated by CreateFlatFiles_Runscript.R, an RMarkdown (.rmd) file that takes files (see Data Used in the App), wrangles the data with a variety of functions, and saves them as .rds files in

\WGFP_dataclean_vis2.0\data\flatFilesforApp. This wrangling is done outside the app to reduce load time and computations within the actual app as much as possible. Anytime there is new data for any of these files, the runscript must be ran in order to update the data in the app.

The typical workflow for updating the app for new data is below, following the use of UpdateDataApp to combine new detections with the previous detection files.

Raw stationary data output saved from updateData app in .rds format goes in the data directory of the data vis app

- Open R project file within the data vis app directory
- Navigate to, open, and Run or Knit CreateFlatFiles Runscript.rmd

Navigate to, open, and run Data Visualization app

- The resulting <u>dataframe</u> is a cleaned file of stationary data
- Biomark process still uses .csv
- 2023 example Stationary file size of .rds file vs .csv: ~13 mb vs ~157 mb
- Uses tabular files outlined in the "Data used in the app" section (File Readins)
- Creates static files used in the data visualization app by cleaning and combining with other files in "data" directory (Creating New Flat Files/Running the Script)
- Outputs .rds files into the flatFilesforApp folder
- Typically takes a little less than 5 minutes
- Needs to be run anytime there is a change to a data file to get the most up to date data in the app; this includes files in the "data" directory and updated spatial .<u>shp</u> files in the "gis" directory
- App loading time cut down by at least 3 minutes
- Output should be consistent without the need to clean the files within the app
- Check to make sure the expected up to date data is included

File Readins

Anytime there is new data of any of the files ending in yyyymmdd listed above, the name of the file must be updated in the Runscript.

Column names and order matter. When updating the csv or excel files, make sure the new file has the same column name/order as the old one.*

Column names outlined below are how they appear when brought into R at the beginning of the app. In Excel, names will appear slightly differently.

*If you do need to change the column names or order, you'll have to go into the app and follow where that column is used in the app and change the name

Make sure tag is read in correctly. The way to do this is by saving the Tag column as numeric with 0 decimal places, instead of general (default). If not, the runscript won't run all the way through (see Common Errors)

Updating file names

To update the file name, change these dates/names for the appropriate file.

```
Stationary <- readRDS("./data/WGFP_Raw 20240514 rds")

Mobile <- read.csv("./data/WGFP_Mobile_Detect_AllData.csv" , colClasses= c(rep rep("character", 3)))

Biomark <- read.csv("./data/Biomark_Raw_20221102.ssv", dec = ",")

# need to have tagID as a numeric field in the .csv file in order to be read in Release <- read.csv("./data/WGFP_ReleaseData_Master_20240520 csv", na.strings colClasses=c(rep("character",8), "numeric", "numeric", rep("character",8)))
```

Other Data Updates

- Pressure Transducer: overwrite the current file with the new one. If a new site is added, add that file to the "data/PressureTransducer" directory. Open "WGPF Metdata.xlsx" (located in the data folder of the app, and U:\Projects\Colorado_River\Windy_Gap_FishMovementStudy\Data\RFID\Detections) and enter the correct "PressureTransducerSiteName" as it corresponds to the name used in the PT file. Names in the PT file need to match the No need to change anything in the app, it will automatically get read in and combined with the other site data as long as it has the same columns/order as the other data
- USGS data: scraped from the USGS server and is updated when the runscript runs.
- All other files: overwrite the previous file, making sure column names/order were the same as the previous one

Creating New Flat Files/Running the Script

To run the script and update the data in the app, click "Knit"

Progress will show in the "Render" window

```
.../WGFP_dataclean_vis2.0/CreateFlatFiles_Runscript.Rmd
| ..... | 13% [readins]

processing file: CreateFlatFiles_Runscript.Rmd
| ..... | 27% [USGSData]
```

At the end, the document will output a HTML document just to confirm that the data was saved. On a local machine this usually takes 3-4 minutes, sometimes up to 10 or more when using a vpn or directly on the U drive. Saving the files takes particularly long on the U drive.

Create Files for App Runscript

```
Sam Graf
2023-07-06

## [1] "Reading in USGS Data took 13.65"

## [1] "Running All_combined_events_function: Combining and cleaning Stationary, Mobile, Biomark, Release, and Rec apture csv inputs....."
## [1] "All_combined_events_function took 1.13 minutes"

## [1] "Running spatial_join_stations_detections function: Joining detections and events to stations shapefile."
## [1] "Spatial_join_stations_detections took 0.03 minutes"

## [1] "Running States Function: Assigns letters A, B, C, or G based on position relative to dam, or Ghost/predate d tag."
## [1] "States Function took 0.26 minutes"

## [1] "Running Ind_tag_enc_hist_wide_summary_function: Summarizes detection and movement data from each released Tag."
## [1] "Encounter Histories Summary Wide Function took 0.06 minutes"

## [1] "Running get_movements_function: Calculates movements of fish based off a change in station."
## [1] "Movements Function took 0.34 minutes"

## [1] "Saving Files took 0.76 minutes"
```

The whole document took 3.27 minutes to run. Flat Files are saved to the flat files directory and the data vis app is ready to run with the most updated data.

There is some QAQC that happens in this runscript as well, including checking if Tags have multiple entries in the Avian Predation, ghost tag, and release files. If they do, it will show up in the final HTML doc. This also catches if tags were saved incorrectly (see Common Errors).

The resulting flat/static files are saved to data/flatFilesforApp and are automatically used in the app, it's ready to use (see How to Open)!

Common errors:

Warning: cannot open file './data/Biomar_Raw_20221102.csv': No such file or directoryError in file(file, "rt"): cannot open the connection.

This usually means the filename is spelled wrong or is in the wrong directory

"Error in scan(file = file, what = what, sep = sep, quote = quote, dec = dec, : scan() expected 'a real', got a

• This typically means that there is an incorrect number of columns in the data compared to what the code was expecting. Check the code where the problem file is being read in to see which column types it expects to find

in each place, and/or modify the excel file. If you aren't sure what a chunk of code does, it's helpful to copy and paste it into chatgpt and it will explain it to you.

Runscript is not able to run all the way through and the console spits out that there are multiple tags of the same name (2.3EII, 2.26EII)

Annoyingly, the tags in one of the csvs didn't save correctly after being opened and closed in excel. In order to
get them to save correctly, the best way is to open the problem file in excel, select the TAG column, change the
type from General to Numeric, move the decimal place over 2 places so there isn't a decimal at all, and resave
it. You may have to go grab the file from the original location on the U drive again is the tag numbers have
already been saved as 230000000000 instead of 230000678123

USGS not being read in

• I can't remember exactly what error this throws but it's something related to server connection. This happens sometimes when the USGS server is not working well and so we're unable to scrape the most recent metrics from their site. Sometimes it helps to not use a VPN, but mainly it's not something we can fix on our end and typically this resolves within a few hours. Otherwise, try to run the script tomorrow.

This section of the app adds dummy rows with fake data from CD, CU, B5, and B6. It was added to make sure the functionality for these new antennas was working, and is not currently in use.

```
source("functions/dummy_rows.R")
dummy_rows_list <- add_dummy_rows(stationary = Stationary, bi
Stationary <- dummy_rows_list$Stationary
Biomark <- dummy_rows_list$Biomark
Release <- dummy_rows_list$Release
ghost_tag_df <- dummy_rows_list$Ghost_tags #date column is na</pre>
```

Modifying, Updating, Adding New Antenna/Stations

There are 2 parts to adding a new antenna to the app: updating the spatial files and updating the code that wrangles the tabular data. You can just update the tabular data code without updating the spatial file if you want.

Updating the code and data that wrangles the tabular data

Metadata

Add new antenna and its data in with WGFP Antenna Metadata (See Data in app)

AntennaSite	SiteName	FrontendSiteCode	BackendSiteCode	PressureTransducerSiteName	DetectionDistance	UTM_X	UTM_Y	River	DeploymentDur-
Biomark test Antenna in random spot	Biomark test Antenna	T1	A5			420727	4437229	Fraser Riv	er

CreateFlatFilesRunscript

Metadata Variable names

- In CreateFlatFilesRunscript, add the variable names to the list in the "metadatavariables" code chunk
 - Do this for a backend variable name and a frontend one. "AntennaSite" column in the metadata is used to identify these codes, so change the code to match what is in AntennaSite

```
"``{r metadatavariables, include=FALSE, echo = FALSE}
###Variables that are used throughout the funcitons
###Frontend (display) codes
TestBiomarkAntennaFrontendSiteCode <-
as.character(wgfpMetadata$AntennaMetadata[!is.na(wgfpMetadata$AntennaMetadata$AntennaSite) &
wgfpMetadata$AntennaMetadata$AntennaSite == "Biomark test Antenna in random spot", "FrontendSiteCode"])
WindyGapBypassAntennaFrontendSiteCode <-
as.character(wgfpMetadata$AntennaMetadata[!is.na(wgfpMetadata$AntennaMetadata$AntennaSite) &
wgfpMetadata$AntennaMetadata$AntennaSite == "Windy Gap Bypass Channel", "FrontendSiteCode"])</pre>
```

Add these variables to metaDataVariableNames

```
metaDataVariableNames <- list(
"TestBiomarkAntennaBackendSiteCode" = restBiomarkAntennaBackendSiteCode,
"TestBiomarkAntennaFrontendSiteCode" = TestBiomarkAntennaFrontendSiteCode.
"WindyGapBypassAntennaFrontendSiteCode" = windyGapBypassAntennaFrontendSiteCode,
"WindyGapAuxiliaryAntennaFrontendSiteCode" = WindyGapAuxiliaryAntennaFrontendSiteCod
"GranbyDiversionAntennaFrontendSiteCode" = GranbyDiversionAntennaFrontendSiteCode,
"RiverRunAntennaFrontendSiteCode" = RiverRunAntennaFrontendSiteCode,
"FraserRiverCanyonAntennaFrontendSiteCode" = FraserRiverCanyonAntennaFrontendSiteCod
```

Dummy Rows

If the antenna doesn't have any new data yet, you will need to set up the functions to account for that antenna. This is done by using the function dummy rows.r

• Navigate to dummy_rows.r (in the functions directory) and add a row to the biomark or stationary data depending on which antenna type we want to add. The only columns that are important are adding the dumy tag number (23000099999) and making sure we've got the right reader ID that lines up with what we put in the metadata

```
Biomark <- biomark1 %>%
    add_row(Scan.Date = "2022-12-07",Scan.Time = "12:52:10.810", Download.Date = Reader.ID = "A3",Antenna.ID = 1,HEX.Tag.ID = "384.358D14F739",
    DEC.Tag.ID = "900.230000999999",Temperature.C = NA,Signal.mV= NA,Is.Duplicate= "Yes",Latitude = NA,Longitude= NA, add_row(Scan.Date = "2022-12-08",Scan.Time = "12:52:10.810", Download.Date = "9/16/2022",Download.Time = "11:30:41",
    Reader.ID = "A4",Antenna.ID = 1,HEX.Tag.ID = "384.358D14F739",
    DEC.Tag.ID = "900.230000999999",Temperature.C = NA,Signal.mV= NA,Is.Duplicate= "Yes",Latitude = NA,Longitude= NA, add_row(Scan.Date = "2022-12-09",Scan.Time = "12:52:10.810", Download.Date = "9/16/2022",Download.Time = "11:30:41",
    Reader.ID = "A5")Antenna.IS = 1,HEX.Tag.ID = "384.358D14F739",
    DEC.Tag.ID = "900.230000999999", Demperature.C = NA,Signal.mV= NA,Is.Duplicate= "Yes",Latitude = NA,Longitude= NA,Longitud
```

- Comment out the lines of code that remove the dummy tag from the data (add them back in when there is actual data for the antenna)
 - In Runscript:

```
355
356  ### taking dummy tag out
357
358  Biomark <- Biomark #%>%
359  #filter(!DEC.Tag.ID %in% c("900.230000999999"))
360  cleanedRelease <- cleanedRelease #%>%
361  #filter(!TagID %in% c("230000999999"))
362  Stationary <- Cleaned_Stationary_FishdetectionsOnly #%>%
363  #filter(!TAG %in% c("900230000999999"))
```

In "get_statesFunction2.r"

```
states <- eventsWithGhostDatesAndAvianPredation %>%
#filter(!TAG %in% c('230000999999')) %>%
mutate(
    #the case_whens apply with priority, so it's important state = case_when(Date >= GhostDate ~ "G",

o In "get_movements_function"

dailyMovementsTable <- combined_events_stations %>%
    #### removing dummy tag
    #filter(!TAG %in% c("230000999999")) %>%
    select(Date, Datetime, TAG, det_type, Event, ET_STATION)
```

o In Ind tag enc hist wide summary function.r:

```
L49

L50  #### dummy rows removal: |

L51  encountersAndRelease6 <- encountersAndRelease6# %>%

L52  #filter(!TAG %in% c("230000999999"))

L53
```

In All combined events.R

```
condensedAllEventsWithReleaseandEnvironmentalInfo <- condensedAllEventsWithReleaseandEnvironmentalInfo
replace_na(list(Species = "No Info", ReleaseSite = "No Info",
Site = "No Site Associated")) %>%
dplyr::filter(!TAG %in% c("23000099999"))|
```

In AllEncountersMod, twice

```
#### filter dummy row

all_events_filtered <- all_events_filtered #%>%

423  #filter(!TAG %in% c("230000999999"))

424

464  output$alleventsfrequencies1 <- renderDataTable({
    frequenciesSummarized <- all_events_data() %>%

466  #filter(!TAG %in% c(230000999999)) %>%

GOURT(Event_name = "Paw Datastions")
```

All combined events function

 In this function, add the following line to the case_when() function based off the Frontend and Backend variable codes

```
biomarkCleaned <- Biomark %>%

mutate(TAG = str_replace(DEC.Tag.ID, "\\.", ""),

# i wish this could be a join, but when there are 2 dif codes (A1, B1, etc) used for backend na

Reader_ID = caco_when(

Reader_ID %in% TestBiomarkAntennaBackendSiteCode ~ TestBiomarkAntennaFrontendSiteCode,

Reader_ID %in% WindyGapBypassAntennaBackendSiteCode ~ WindyGapBypassAntennaFrontendSiteCode,
```

Ind tag hist summary wide function:

Add in your antennaFrontend Variable in column order

```
#column order is just nice to have for the user
columnOrder <- c(RedBarnFrontendCodes, HitchingPostFront
MobileRunFrontendCodes, WindyGapBypassA
GranbyDiversionAntennaFrontendSiteCode,
TestBiomarkAntennaFrontendSiteCode)
```

 Add the frontend site code to these lines of code. If it's a biomark antenna, add it to "TotalBiomark", and if its a Stationary antenna, add to TotalStatiionary

```
57
                    mutate(
58
                        59
                                                                                                                                        ConnectivityChannelDownstreamFrontendCod
60
                                                                                                                                       ConnectivityChannelUpstreamFrontendCodes
61
                                                                                                                                        WindyGapBypassAntennaFrontendSiteCode, W
                                                                                                                                        RiverRunAntennaFrontendSiteCode, FraserR
62
63
                                                                                                                                        TestBiomarkAntennaFrontendSiteCode))) ==
                        TotalAntennas = rowSums(select(., all_of(c(RedBarnFrontendCodes, HitchingPostFrontendCodes, Hitching
64
65
                                                                                                                                  ConnectivityChannelDownstreamFrontendCodes
66
                                                                                                                                  ConnectivityChannelUpstreamFrontendCodes,
67
                                                                                                                                  WindyGapBypassAntennaFrontendSiteCode, Win
                                                                                                                                   GranbyDiversionAntennaFrontendSiteCode,
68
69
                                                                                                                                   RiverRunAntennaFrontendSiteCode, FraserRiv
70
                                                                                                                                  TestBiomarkAntennaFrontendSiteCode))) == T
                        TotalStationary = rowSums(select(., all_of(c(RedBarnFrontendCodes, HitchingPostFronte
71
                                                                                                                                        ConnectivityChannelDownstreamFrontendCod
72
73
                                                                                                                                       ConnectivityChannelUpstreamFrontendCodes
74
                        TotalMobile = rowSums(select(., all_of(MobileRunFrontendCodes)) == TRUE),
75
76
                        TotalBiomark = rowSums(select(., all_of(c(WindyGapBypassAntennaFrontendSiteCode, Wind
77
                                                                                                                                GranbyDiversionAntennaFrontendSiteCode,
                                                                                                                                RiverRunAntennaFrontendSiteCode, FraserRive
78
79
                                                                                                                                TestBiomarkAntennaFrontendSiteCode))) == TR
                        TotalRedBarn = rowSums(select(., all_of(RedBarnFrontendCodes)) == TRUE),
80
81
                        TotalHitchingPost = rowSums(select(.. all of(HitchingPostFrontendCodes)) == TRUE).
```

With that, you are ready to Run the CreateFlatFilesRunscript. Once the data is updated, run the app. Check to make sure your antenna has been integrated correctly in the data by filtering for the dummy tag in the Ecnounters dataframes, movements, and States tabs where it will also have dummy Release Data.

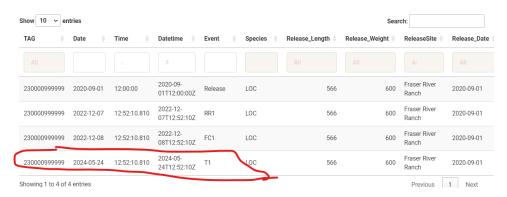


Figure I all Encounters History with Dummy Data



Figure 2 column was automatically created in All EncountersHistory SummaryWide for that new antenna

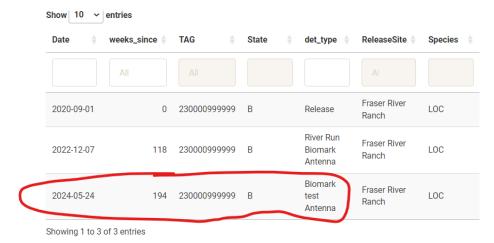


Figure 3States of with dummy data

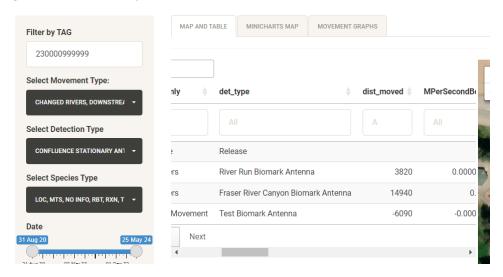


Figure 4 Movements table with dummy data

Adding or updating layers on the map

- In GIS, go to the GIS database and export desired layer as shapefile, saving it in "gis" folder within the app directory
 - Right click on layer in left column, scroll down to data -> export
- Using the sf package, read in the .shp file with the name of your layer

```
layerLocation <- file.path("./gis/")
latLongCRS <- st_crs("+proj=longlat +datum=WGS84 +no_defs") #should be same as +init=epsg:4326
antenna_sites <- st_transform(read_sf(file.path(layerLocation, "antenna_sites1.shp")), latLongCRS)</pre>
```

- The espg:4326 converts the projection to lat/longs able to be plotted with leaflet
- If the layer is very large, it might be good to convert the .shp file to a .rds file, which decreases resolution. For example, the stations file is a .rds file and was converted below. It keeps 10% of the original resolution

```
stations_10m <- readOGR(dsn = layer_location, layer = "stations_10m")
stations_10m <- sp::spTransform(stations_10m, CRS("+init=epsg:4326"))
$imple_stations1 <- ms_simplify(stations_10m, keep = .1)
write_rds(simple_stations1, file = file.path(paste0(layer_location,"/"), paste0("simple_stations.rds")))</pre>
```

• It is read in like so and is much faster than when read in as a shapefile

```
simple_stations2 <- read_rds(file.path("./gis/simple_stations.rds"))
```

- If you make a new layer (not just updating an existing antenna, stations or stream centerline layer that is already included on the map) that you want to display, you'll have to add it to the code that makes the map, found in movementsMod.R.
- Add_polylines is used to bring in lines like stream_centerline, addMarkers is used for SpatialPoints, addPolygons is used for polygons.
- Add a Group argument to the data and add it in the addLayersControl function to make sure it shows up

Adding New Antenna to map

The data will still be displayed spatially on the movements map even without an updated antenna file, but if you want to display the antenna on the map (and you should), then here a couple options to do this. The first option is to add it in GIS, then export the layer (see above) and read in that later.

The second option is to modify the already existing Antenna layer within R. This can be done with already-written code in map polygon readins. R starting on line 57.

Modify the UTMs and other relevant fields to fit the antenna.

```
57
   #####adding new antenna to data
58
   utm_x <- 420727
59
60
   utm_y <- 4437229
    new_point <- st_sfc(st_point(c(utm_x, utm_y)), crs = 32613) # EPSG:32613 is UTM zone 13N</pre>
61
62
    # Transform the new point to the desired CRS (latLongCRS)
63
    new_point_transformed <- st_transform(new_point, st_crs(latLongCRS))</pre>
64
65
66
   new_row <- tibble(</pre>
67
     OBJECTID = NA,
      SiteName = NA,
68
      Objective = NA,
69
70
      StudyPerio = NA,
      SiteLabel = NA,
71
      AntennaNam = NA,
72
73
      WaterName = NA,
74
      ChannelWid = NA,
75
      UTM_X = utm_x,
76
      UTM_Y = utm_y,
77
      Notes = NA,
      geometry = new_point_transformed
78
79 )
```

 Add this row to the data and save it in the correct location with st_write. This WILL overwrite the previous file.

```
antenna_sites1 <- antenna_sites %>%
add_row(new_row)

output_file <- file.path(layerLocation, "antenna_sites1.shp")

# Write the updated sf object to a new shapefile
st_write(antenna_sites, output_file)</pre>
```

• Once the new layer is saved, comment out those lines of code from lines 57 to 87 to make sure these lines aren't ran again.

Functions

Here are some of the main functions of the app.

If you don't know what a chunk of code does, it is helpful to copy and paste the chunk into chatGPT it it does a good job explaining it.

Note: some variables used in the functions are not included as arguments, but are read in earlier in the Runscript so should always exist before running.

All Combined events function

- Main function that combines and cleans detections and release files
- o Takes Stationary, Mobile, Biomark, Release, and Recapture files as arguments
- o **Returns**:
 - WGFP Clean: a clean dataset of all stationary detections, no weird tags or marker tags, or duplicate rows. All timestamps and dates are in the same format. 900_ is taken off of the tags
 - All_Events: cleaned dataset of all 5 input datasets, containing every detection/event from each tag that hit an antenna, complete with release info.
 - Marker Tag data: file of just marker tags, to be used in QAQC
 - Recaps_detections: a file of all antenna detections and recaptures, but no release data.
 This is just to be used in enc_hist_wide_summary_function

About the Code

The app first splits Stationary dataset into a dataset with no marker tags, and one with only marker tags. It also removes known test tags

The Marker tags are cleaned, first putting them into a mdy date format.

Then it goes through a timestamp cleaning process that is necessary because there are timestamps that have AM/PM, and others that do not. So we need to get them to military time.

```
Markers_only1 <- Markers_only %>%
  mutate(
    DTY =
      ifelse(str_detect(DTY, "/"),
                 as.character(mdy(DTY)),
                 DTY),
    DTY2 = as.Date(DTY))
Markers_only2 <- Markers_only1 %>%
  #this is the same process that all_detections goes through
                                                   yM") & str_detect(ARR, "^12:") ~ hms(ARR) - hours(12),
PM") & str_detect(ARR, "^12:") ~ hms(ARR),
  str_detect(ARR, "AM") & str_detect(ARR, "^12:",
str_detect(ARR, "PM") & str_detect(ARR, "^12:",
negate = TRUE) ~ hms(ARR) + hours(12),
                                #if it doesn't detect PM or AM just do hms(ARR)
                                str_detect(ARR, "PM|AM") == FALSE ~ hms(ARR)),
         Scan_Time2 = as.character(as_datetime(Scan_Time1)),
         CleanARR = str_trim(str_sub(Scan_Time2, start = 11, end = -1))
  select(Code, DTY2, ARR, CleanARR, TRF, DUR, TTY, TAG, SCD, ANT, NCD, EFA) %>%
  rename(DTY = DTY2)
```

Then the rest of the stationary data without marker tags or test tags is cleaned, including changing antenna names for the upcoming multiplexor, formatting dates, assigning UTMs for each antenna, and removing duplicate rows if there are any

```
## rest of getting "clean" windy gap stationary data
### Subset Detection Type "Codes" to only include Summary (S) and Individual (I) ###
WGFP_Clean= data.frame(WGFP_NOMarkers[which(WGFP_NOMarkers$Code == "I" | WGFP_NOMarkers$Code == "S"),])
#### Add Lat Longs to detections ###
# takes out 900 from TAG in WGFP Clean
# also takes out duplicate rows
WGFP_clean <- WGFP_clean %>%
  mutate(TAG = ifelse(str_detect(TAG, "^900"), str_sub(TAG, 4,-1), TAG), SCD = case_when(SCD == "CD7" & ANT == "A1" ~ "CD7", SCD == "CD7" & ANT == "A2" ~ "CD8",
                              SCD == "CD7" & ANT == "A3" ~ "CD9"
                              SCD == "CD7" & ANT == "A4" ~ "CD10",
                             TRUE ~ SCD),
          DTY = ifelse(str_detect(DTY,
                           as.character(mdy(DTY)),
                           DTY)) %>%
  # assigning UTM's are important because they are plotted later when getting stations file in GIS
  mutate(UTM_X =case_when(SCD == "RB1" | SCD == "RB2" ~ "412489", SCD == "HP3" | SCD == "HP4" ~ "414375", SCD == "CF5" | SCD == "CF6" ~ "416965",
                               SCD == "CD7" | SCD == "CD8" | SCD == "CD9" | SCD == "CD10" ~ "415801",
                               SCD == "CU11" | SCD == "CU12" ~ "416802"),
                                               | SCD == "RB2" ~ "4439413"
| SCD == "HP4" ~ "4440241"
| SCD == "CF6" ~ "4439369"
          UTM_Y = case_when(SCD == "RB1"
                                SCD == "HP3"
                                SCD == "CF5"
                                SCD == "CD7" | SCD == "CD8" | SCD == "CD9" | SCD == "CD10" ~ "4439899",
                                SCD == "CU11" | SCD == "CU12" ~ "4439507")) %>%
  distinct()
```

Biomark dataset is then cleaned, including changing the reader ID since we want it in B3-B6 format, changing date format, filtering out test tags and marker tags, assigning UTMs, and removing duplicate rows

```
# biomark cleaning, getting dates into uniform format,
biomark2 <- Biomark %>%
 mutate(TAG = str_replace(DEC.Tag.ID, "\\.", ""),
         Reader.ID = case_when(Reader.ID == "A1" | Reader.ID == "B1" ~ "B3",
                               Reader.ID == "A2" | Reader.ID == "B2" ~ "B4"
                               Reader.ID == "A3" ~ "B5",
                               Reader.ID == "A4" ~ "B6",
                               TRUE ~ Reader.ID),
         #make a column for Scan>Date if parentheses are detected in the string,
         # and we want to convert it to YYYYMMDD format. elsewise, leave it as i
         Scan.Date = ifelse(str_detect(Scan.Date, "/"),
                            as.character(mdy(Scan.Date)),
                            Scan. Date)
           ) %>%
 filter(!TAG %in% c("900230000102751", "900226001581072", "999000000007586", "9
 # from gis: B1 416026, 4440196
 #B2: 420727.9, 4437221
 # b3 is wg, b4 is kaibab
 # b5 is river run
  # b6 is <u>fraser</u> river canyon
 mutate(UTM_X =case_when(Reader.ID == "B3" ~ "416026",
                          Reader.ID == "B4" ~ "420728",
                          Reader.ID == "B5" ~ "419210",
                          Reader.ID == "B6" ~ "424543")
         UTM_Y = case_when(Reader.ID == "B3" ~ "4440196"
                           Reader.ID == "B4" ~ "4437221",
                           Reader.ID == "B5" ~ "4439961'
                           Reader.ID == "B6" ~ "4435559")) %>%
 distinct()
```

Then Mobile, Stationary, and Biomark columns are all renamed and then they are joined together. Mobile data requires a little Tag cleaning but not much.

```
###Create one big clean dataset
WGFP_condensed <- WGFP_clean %>%
 select(DTY, ARR, TAG, SCD, UTM_X, UTM_Y) %>%
 rename(Scan_Date = DTY, Scan_Time = ARR, Site_Code = SCD, UTM_X = UTM_X, UTM_Y = UTM_Y)
Biomark_condensed <- biomark2 %>%
  mutate(TAG = ifelse(str_detect(TAG, "^900"), str_sub(TAG, 4,-1), TAG)) %>%
  select(Scan.Date, Scan.Time, TAG, Reader.ID, UTM_X, UTM_Y) %>%
 rename(Scan_Date = Scan.Date, Scan_Time = Scan.Time, Site_Code = Reader.ID, UTM_X = UTM_X, UTM_Y = UTM_Y)
Mobile_condensed <- Mobile %>% #gonna have to just change to mobile eventually
  rename(TAG = TagID) %>%
  mutate(TAG = ifelse(str_detect(TAG, "^900"), str_sub(TAG, 4,-1), TAG),
         Date = ifelse(str_detect(Date, "/"),
                       as.character(mdy(Date)),
                       Date)) %>% #end of mutate
  select(Date, Time, TAG, Ant, UTM_X, UTM_Y) %>%
  rename(Scan_Date = Date, Scan_Time = Time, Site_Code = Ant)
WG_bio <- bind_rows(WGFP_condensed, Biomark_condensed)
All_detections <- bind_rows(WG_bio, Mobile_condensed)
```

Scan time is now cleaned like it was for marker tags, but with all the data

Data is filtered to only include detections past the study start date, and datetime is put in a good format

```
All_detections2 <- All_detections1 %>%
filter(Scan_Date >= as.Date("2020-08-06")) %>% #right before the first date of mutate(
   #datetime1 = as.POSIXct(paste(Scan_Date, Scan_Time),format="%Y-%m-%d %H:%M:%S Scan_DateTime = ymd_hms(paste(Scan_Date, clean_time))) %>%
#rename(Scan_DateTime = datetime2) %>%
select(Scan_Date, clean_time, Scan_DateTime, TAG, Site_Code, UTM_X, UTM_Y)|
```

Release and recapture files are brought in and timestamps are cleaned. There is need for this because sometimes the way that times are entered in the spreadsheet are not uniform. Columns are renamed for joining.

```
### all detections and recaps and release "EVENTS" DF
  #getting timestamps in order and getting relevant columns
  Release1 <- Release %>%
    rename(TAG = TagID) %>%
    mutate(TAG = str_trim(TAG),
           Date = mdy(Date),
           Time1 = case_when(str_length(Time) > 5 ~ as_datetime(hms(Time)),
                               str\_length(Time) <= 5 \sim as\_datetime(hm(Time))), #warning message: problem with mutate(time1,
           Time2 = str\_sub(Time1, start = 11, end = -1),
           DateTime = ymd_hms(paste(Date, Time2))) %>%
    select(RS_Num,River,ReleaseSite,Date, Time2, DateTime,UTM_X,UTM_Y,Species,Length,Weight,TAG,TagSize,Ant,Event) %>%
    rename(Time = Time2)
  #getting timestamps in order and getting relevant columns
  recaps1 <- Recaptures %>%
    rename(TAG = TagID) %>%
filter(!Date %in% c("", " ", NA)) %>%
    mutate(TAG = str_trim(TAG),
           Date = mdy(Date),
            # added in like I did the release file, functionality for when the time contains just HH:mm and hh:mm:ss
           \label{time1} \mbox{Time1} = \mbox{case\_when(str\_length(Time)} \ > \ 5 \ \sim \mbox{ as\_datetime(hms(Time))} \,,
                               str_length(Time) <= 5 ~ as_datetime(hm(Time))),</pre>
           Time2 = str\_sub(Time1, start = 11, end = -1),
            DateTime = ymd_hms(paste(Date, Time2))) %>%
    select(RS_Num,River,RecaptureSite,DateTime,Date,Time2,UTM_X,UTM_Y,Species,Length,Weight,TAG,TagSize,Ant,Event) %>%
    rename(Time = Time2,
           Recap_Length = Length,
           Recap_Weight = Weight
    )
```

Reformatting the all detections file to also be ready to join, then binding release and recap data to the df. Binding the release df to the main df makes it so there will be an event for "release", and then left joining will then give release info for each fish for each detection.

Then there's some minor formatting/renaming for display purposes in the app, and this is the final df that is used in the All Encounters Tab!

```
#Change na to "No info" in select columns so that it will register with the Picker input in the app
#pretty sure that's just a bug on the DT or <u>shinyWidgets</u> end that it can't select by NA
# 87 rows were not even showing up on the all_events app because the Species was NA -12/14/21 SG
filled_in_release_rows_condensed <- filled_in_release_rows %>%
  select(Date.x, Time.x, DateTime.x, TAG, Event.x, Species.y, Length.y, Weight.y, ReleaseSite.y, Date
  rename(Release_Date = Date.y,
         Date = Date.x,
         Time = Time.x,
         Datetime = DateTime.x,
         Event = Event.x,
         Species = Species.y,
         Release_Length = Length.y,
         Release_Weight = Weight.y,
         ReleaseSite = ReleaseSite.y,
         UTM_X = UTM_X.x,
         \mathsf{UTM}\_\mathsf{Y} \; = \; \mathsf{UTM}\_\mathsf{Y.\,X}) \; \; \% \!\!> \!\! \%
  #gets rid of all duplicate rows but keeps all info
  distinct(Datetime,TAG, Event, .keep_all = TRUE) %>%
  replace_na(list(Species = "No Info", ReleaseSite = "No Info"))
```

This line makes sure that the tags displayed are from the release file. There is a tab in the QAQC tab to see unknown tags histories.

```
#makes sure all events are from tags ONLY in the release file
filled_in_release_rows_condensed <- left_join(Tags_only, filled_in_release_rows_condensed, by = "TAG")</pre>
```

This data frame condenses detections down to their daily summary; so it will take the first detection a fish had that day, and the last detection a fish had that day, and all unique detections in between. This reduces a fish to the "most relevant" detections. This df is ultimately used in the movements df, but first in joining with the Stations Data, which is the next function. WARNING: It's important to note that if a fish has a day where the movement sequence is like "RBI, RB2, HP4, HP3, HP4, RB2, HP3, HP4", then for that day, the sequence will register as "RBI, RB2, HP4, HP3, HP4".

Spatial join function

- Joins detections and events to the shapefile of stations in order to later help calculate states and distance moved for each fish.
- Takes condensed events (arg I), made from all_combined_events_function
 (all_events_most_relevant). Also takes simple_stations (arg2), which is usually read in as simple_stations2 in the polygon_read_ins.r file
- o Returns a dataframe of condensed events with all station data attached.

About the Code

This function first converts the utm's to numeric, assigns a projection/zone/datum, converts to lat/long, then to a spatial points object, then to sf object, then joins with the stations shapefile, also converted to an sf object. The joining is by the nearest feature, so if a detection is not quite on top of the station, it will get assigned the closest one.

```
condensed_events <- condensed_events %>%
     X = as.numeric(UTM_X),
     Y = as.numeric(UTM_Y)
  ) #end of mutate
# assigning projection to ready \frac{df}{df} = \frac{1}{13}" attr(condensed_events, "zone") = "13" attr(condensed_events, "projection") = "UTM" attr(condensed_events, "datum") = "GRS80"
# need a column that has x and Y for this
# converts lutms to lat/long
condensed_events <- convUL(condensed_events, km=FALSE, southern=NULL)</pre>
#converting lat/long entries to spatial points dataframe
# needs to have a df of just coordinates (xy)
xy <- condensed_events %>%
  select(X, Y)
spdf <- SpatialPointsDataFrame(coords = xy, data = condensed_events,
                                      proj4string = CRS("+init=epsg:4326"))
## making sf objects
detections_sf <- st_as_sf(spdf)
stations_sf <- st_as_sf(simple_stations)
joined <- st_join(detections_sf, stations_sf, st_nearest_feature)</pre>
#need to convert class sf object back to <u>dataframe</u> so that it goes faster in combine_events_stations_function
station_data <- as.data.frame(joined)
```

PrepareforStatesMovementsandSummary function.R

Puts stations onto a condensed All events dataframe

- Takes StationData returned from spatial join function.
- Returns:
 - All_events_days I: a condensed dataframe of all pertinent daily detection info for each tag with stations attached.
 - Row entries for tags with multiple detections on the same UTM_X, UTM_Y, at
 the same antenna, on the same day are filtered out. Leaving the first and last
 detections of the day, and all detections on unique antennas and UTM's in
 between.

About the Code

The columns are renamed and the River is assigned for specific antennas

```
mutate(
  #River also needs to be assigned for new detections
River = case_when(
    (Event %in% c("RB1", "RB2")) ~ "Colorado River", # there is no is.na here because RB UTM
    (Event %in% c("HP3", "HP4")) ~ "Colorado River",
    (Event %in% c("CF5", "CF6")) ~ "Colorado River",
    (Event %in% c("CD7", "CD8", "CD9", "CD10", "CU11", "CU12")) ~ "Connectivity Channel",
    (Event %in% c("B3", "B5")) ~ "Colorado River",
    (Event %in% c("B4", "B6")) ~ "Fraser River",
    TRUE ~ River
),
```

This part accounts for stations up the Fraser River, because those stations start at 0. This adds the station where the Colorado/Fraser confluence to the Fraser stationing. Then duplicate rows are taken out (shouldn't be any) and relevant columns are selected

This part gets the number of daily unique events and detections for a fish, lets you know if a detection was above and below the dam. The type of movement is condensed to a "detection type" field. This is helpful in the movements map where it doesn't necessarily matter which antenna specifically is hit, only if the stationing is different between events.

Ind tag enc hist wide summary function

- Makes a summary dataframe of all released/tagged fish with over 60 columns of summary info, like "total number of antenna encounters", "total distance travelled", "moved through dam or not"
- Takes recaps_and_all_detections from All_Combined_events_function, Release data, combined_events_stationsfrom PrepareforStatesMovementsandSummary_function, and States summarized from the Get states function
- It doesn't take All Events from All_combined_Events_function because we want to do some operations on the events that aren't Releases, before bringing the release data back in for joining.
- o Returns:
 - ENC_Release_wide_summary : Summary "wide" dataframe of each tagged fish and summary info
 - Unknown_Tags: dataframe of tags that have encounters, but have no release info

About the Code

This part counts the number of Events for each fish. It then pivots the data to wide format, so each fish/tag has a row. Columns are then renamed. This is the start of the summary file.

```
all_enc12 <- recaps_and_all_detections %>%
  count(TAG, Event, name = "Encounters")
all_enc12 <- pivot_wider(data = all_enc12, id_cols = TAG, names_from = Event, values_from = Encounters)
x <- all_enc12 %>%
 replace_na(list(Species = "No Info", ReleaseSite = "No Info"))
#all_enc12[is.na(all_enc12)]=0
#### NEED To make this compatible with new antennas!!
ENC_ALL <- all_enc12 %>%
 rename(RB1_n = RB1,
        RB2_n = RB2,
        HP3_n = HP3,
        HP4_n = HP4
        CF5_n = CF5
        CF6_n = CF6,
        CD7_n = CD7,
        CD8_n = CD8,
        CD9_n = CD9
        CD10_n = CD10,
        CU11_n = CU11,
        CU12_n = CU12.
        M1_n = M1,
        M2_n = M2
        B3_n = B3,
         B4_n = B4,
        B5_n = B5,
        B6_n = B6,
         Recap_n = Recapture
  ) %>%
  select(TAG, RB1_n,RB2_n,HP3_n, HP4_n, CF5_n, CF6_n, CD7_n, CD8_n, CD9_n, CD10_n, CU11_n, CU12_n, M1_n,
```

The release data is then joined to the summary file. This is also when the unknown tags df is made (displayed later in qaqc). These tags all start with 900. They are probably cormorants or mergansers that have swallowed other PIT tagged trout and chubs from glenwood;)

```
#### Combine Release data ###
Release1 <- release_data %>%
    rename(TAG = TagID) %>%
    mutate(TAG = str_trim(TAG)) %>%
    replace_na(list(Species = "No Info", ReleaseSite = "No Info")
# was geting a massive dataframe because the Release df is ca #need to actually join on full join not merge
ENC_Release <- full_join(Release1, ENC_ALL, by = "TAG")
#gets tag list that wasn't in release file
unknown_tags <- ENC_Release %>%
    filter(is.na(ReleaseSite)) %>%
    select(TAG,where(is.numeric))
```

If there wasn't a count for a fish, that NA gets changed to 0. True/falsecolumns are made on whether a fish hit a specific antenna or was recaptured.

```
#ENC_Release11$TAG[3433:nrow(ENC_Release11)]
ENC_Release[is.na(ENC_Release)]=0 #gets rest of the number count columns to 0 from NA
#### Make 1 or 0 for encounter history rather than counts ###
#gets of with TF of whether a fish was detected at a antenna
ENC_Release1 <- ENC_Release %>%
 mutate(RB1 = (RB1_n > 0),
         RB2 = (RB2_n > 0),
         HP3 = (HP3_n > 0),
         HP4 = (HP4\_n > 0),
         CF5 = (CF5_n > 0),
         CF6 = (CF6_n > 0),
         CD7 = (CD7_n > 0),
         CD8 = (CD8_n > 0),
         CD9 = (CD9_n > 0),
         CD10 = (CD10_n > 0),
         CU11 = (CU11_n > 0),
         CU12 = (CU12_n > 0),
         M1 = (M1_n > 0),
         M2 = (M2_n > 0),
         B3 = (B3_n > 0),
         B4 = (B4_n>0),
         B5 = (B5_n > 0),
         B6 = (B6_n>0),
         Recapture = (Recap_n > 0)
```

Here some summary statistics based on the previous columns are calculated. It tells the number of detections a fish may have at a paired antenna. Also a T/F column to see if a fish was detected at a paired antenna. The fish with unknown release data are also filtered out here.

```
totalcols <- ncol(ENC_Release1)
ENC_Release2 <- ENC_Release1 %>%
  #counts number of TRUE across specified rows. -SG
  # need to have parentheses (totalcols-1) that's why i was getting bad numbers on
  #added 8 new columns for new antennas
  mutate(
    TotalEncounters = rowSums(ENC_Release1[(totalcols-18):totalcols] == TRUE),
    TotalAntennas1 = rowSums(ENC_Release1[(totalcols-18):(totalcols-1)] == TRUE),
    TotalStationary = rowSums(ENC_Release1[(totalcols-18):(totalcols-7)] == TRUE),
    TotalMobile = rowSums(ENC_Release1[(totalcols-6):(totalcols-5)] == TRUE),
    TotalBiomark = rowSums(ENC_Release1[(totalcols-4):(totalcols-1)] == TRUE),
    TotalRB = rowSums(ENC_Release1[(totalcols-18):(totalcols-17)] == TRUE),
    TotalHP = rowSums(ENC_Release1[(totalcols-16):(totalcols-15)] == TRUE),
    TotalCF = rowSums(ENC_Release1[(totalcols-14):(totalcols-13)] == TRUE),
    TotalCD = rowSums(ENC_Release1[(totalcols-12):(totalcols-9)] == TRUE),
    TotalCU = rowSums(ENC_Release1[(totalcols-8):(totalcols-7)] == TRUE),
  # just says if the fish was ever detected at these sites
  mutate(RB = (RB1_n > 0 \mid RB2_n > 0),
         HP = (HP3_n > 0 \mid HP4_n > 0),
         CF = (CF5_n > 0 \mid CF6_n > 0),
         CD = (CD7_n > 0 \mid CD8_n > 0 \mid CD9_n > 0 \mid CD10_n > 0),
         CU = (CU11_n > 0 \mid CU12_n > 0),
         Biomark = (B3_n > 0 \mid B4_n > 0 \mid B5_n > 0 \mid B6_n > 0),
         Mobile = (M1_n > 0 \mid M2_n > 0)) %>%
  filter(!UTM_X %in% c(0, NA)) # one way to filter out tags that don't have any son
```

This part brings the condensed df with stations back in for summaries. It first counts the number of encounters each fish had above and below the dam. Then describes that event (eg "Mobile Detection above the dam"). Then if pivots the data wider to get each fish its own row and get in a format for joining back to the original release summary file. It also changes the NA's to FALSE, because when you use count(), you only sum fish that have had encounters.

The data is joined with the original summary file. A column is made based on those new joined columns describing if a fish went through the dam or not.

The states_summarized dataframe is then brought in and added. This has data about how a fish got through the dam, if it used the connectivity channel or not.

```
# left joining states summary to enc_release
ENC_Release5 <- left_join(ENC_Release4, States_summarized, by = "TAG")
#rearranging so that Tag is first column shown
ENC_Release5<- ENC_Release5 %>%
   select(TAG, 1:ncol(ENC_Release5))
```

The distance a fished moved is then calculated from the condensed df detections (see the All Events Combined description to see how this is formed). Here, some data might be lost since it is using the condensed "most relevant" daily rather than every single detection, but it's pretty rare that that will matter anyway, because most fish that that warning/edge case would apply to have actually been predated.

The total movement is calculated for each fish in the sequence of its encounter history. This is then joined to the summary dataframe as well.

Dummy rows are also taken out

```
###joining on column with sum data
#same code appears in movements function
sum_dist1 <- combined_events_stations %>%
 group_by(TAG) %>%
 arrange(Datetime) %>%
 mutate(dist_moved = ET_STATION - lag(ET_STATION, order_by = Datetime),
        sum_dist = (sum(abs(diff(ET_STATION, na.rm = TRUE))))
 ) %>% #end of mutate
 distinct(TAG, .keep_all = TRUE) %>%
 select(TAG, sum_dist)
ENC_Release6 <- ENC_Release5 %>%
  left_join(sum_dist1, by = "TAG")
#### dummy rows removal: 1/14/23
ENC_Release6 <- ENC_Release6 %>%
 filter(!TAG %in% c("230000999999"))
```

Get movements function

- o Makes a condensed dataframe of fish "movements"; see "movements defined" for more info
 - A movement is defined as a change in stationing (assigned to detections in the spatial_join function). A fish that continuously hits RBI and RB2 for example would register as a "No Movement", but a fish that goes from RB2 to HP3 would register as an "Upstream Movement" on HP3. All data is incorporated in this calculation, so if a fish was released below red barn, hit the red barn stationary antenna, then was detected by a mobile run upstream of red barn, then was recaptured downstream of that mobile detection, it will register as a "Initial release", "Upstream Movement", "Upstream Movement", "Downstream Movement". The absolute total of this distance is also summed and displayed as a column in the Encounter Histories Summaries Wide tab.
- o Takes all events days I from PrepareforStatesMovementsandSummary function
- Returns:
 - Movement_table_no_trans: dataframe where only daily movements are included per tagged fish on unique antennas and UTM's. See "Movements defined" for about them

About the Code

The code first groups by each tag and arranges each tag in order by their first, then subsequent events. It then calculates the distance moved between each event for each fish, based on the stationing. The code accounts for Fraser/Colorado River transitions by first calculating the distance to the confluence from the last event, then finding the distance from the confluence to the present event, then adding those together. Otherwise, it's just the difference between the last event and the current one.

The sum distance is then calculated as an absolute value of each movement that was calculated for a fish.

Marker Colors and icon colors are then assigned for later mapping.

Lastly in this code chunk, X and Y are defined as columns in preparation for changing the utms to lat/longs. Note: there are a couple places in the app where UTMs are converted to lat/longs, and the code would be a little more elegant if these were just all made to sf objects, but this stuff also all works fine.

```
movement table notrans <- combined events stations %>%
  #### removing dummy tag
filter(!TAG %in% c("230000999999")) %>%
  select(Date, Datetime, TAG, det_type, Event, ET_STATION, Species, Release_Length, Release_Weight, ReleaseSite, Release_Date, Recaptur
  #grouping by TAG and arranging by datetime makes sure that total distance moved is totalled and summed in order
  group_by(TAG) %>%
  arrange(Datetime) %>%
  #dist_moved would be the place to fenaple new movements based on previous event...je hitting wg biomark followed by connectivity chan
  #trickier but doable for mobile runs
### accounting for FRASER/UPPER COLROADO MOVEMENTS
  if previous station is above the confluence and current station is above the confluence and you changed rivers,
  #then take the previous station and subtract the confluence station to get distance travelled to the confluence (A), then subtract th
  # otherwise, just subtract current station from previous
  mutate(dist_moved = case_when(lag(ET_STATION, order_by = Datetime) > 10120 & ET_STATION >10120 & River != lag(River, order_by = Datet

TRUE ~ ET_STATION - lag(ET_STATION, order_by = Datetime)
         sum_dist = (sum(abs(dist_moved), na.rm = TRUE)),
         movement_only = case_when(lag(ET_STATION, order_by = Datetime) > 10120 & ET_STATION >10120 & River != lag(River, order_by = Da Event %in% c("Release", "Recapture and Release") ~ "Initial Release",
                                     Event %in% c("Release", "Recapture a
dist_moved == 0 ~ "No Movement",
dist_moved > 0 ~ "Upstream Movement"
                                     dist_moved < 0 ~ "Downstream Movement"),</pre>
         #this is for mapping later on
         movement_only == "Initial Release" ~ "orange
movement_only == "Changed Rivers" ~ "purple"
                                                                           "orange",
                                    #str_detect(movement_only, "Initial Release (or recapture and release)") ~ "yellow"
         X = as.numeric(UTM_X).
         Y = as.numeric(UTM_Y)
  ) #end of mutate
```

The UTM's are fully converted to Lat/longs, then the df is parred down with the distinct() function like it has before, but this time, it takes "movement" into account instead of "first/last". Movement_table_notransitions I is what is mapped and filtered in the app.

```
# assigning projection to ready df lat/longs for plotting
attr(movement_table_notrans, "zone") = "13"
attr(movement_table_notrans, "projection") = "UTM"
attr(movement_table_notrans, "datum") = "GRS80"

# need a column that has x and Y for this
# converts lutms to lat/long
movement_table_notrans <- convUL(movement_table_notrans, km=FALSE, southern=NULL)
#as of now, movement table still has rows reminiscent from first_last etc which are helpful when you want to know wher
#but if you want to know concise movments, then this will eliminate uneeded rows
#example: 230000142723
movement_table_notrans1 <- movement_table_notrans %>%
    distinct(Date, TAG, det_type, movement_only, UTM_X, UTM_Y, .keep_all = TRUE) %>%
    select(Date, Datetime, TAG, movement_only, det_type, dist_moved, sum_dist, ET_STATION, Species, Release_Length, Rele
#giving id column to make map proxy easier
# actually the id column needs to be remade every time a filter is applied. See the movements_df_reactives
#movement_table_notrans1$id <- seq.int(nrow(movement_table_notrans1))</pre>
```

Get States function

- Makes dataframe of weekly "states" either above or below the dam. Below the dam is state A
 and above is B. State C is a detection in the connectivity channel. Ghost state is G, and
 Predated state is P
 - Any detection, recapture, or release is registered as a state
 - A fish is inferred to remain in that state until it becomes a ghost tag or is detected/recaptured again.
- Takes combined_events_stations from PrepareforStatesMovementsandSummary_function, the Ghost tags df, and the aviation predation df.

Returns:

- All_States: Dataframe of daily States of fish. Certain assumptions are made to infer transitions, such as if a fish hits 1/2 antennas at a site then hits an upstream antenna, it is said that an upstream transition originally occurred and it just missed 1 antenna. For more assumptions see thee Daily States Tab section
- Days_and_states_wide: same data as all_states but in wide format where days are the columns, TAG is the first column, and values are states
- Flagged Movements: States that couldn't be accounted for in the code

About the Code

The 3 dataframes are joined then the df is cut down to the most succinct: rows with a distinct combo of Tag, Event, Datetime, UTM's, and first and last event of day.

```
# these dates are cleaned before they go into this function
ghost_tag_df <- GhostTags %>%
    rename(TAG = TagID) %>%
    select(TAG, GhostDate)

av_pred_df <- AvianPredation %>%
    rename(TAG = TagID) %>%
    select(TAG, PredationDate)

#joining with ghost tag df
wghost_date <- left_join(combined_events_stations, ghost_tag_df, by = c("TAG"))
# joining with avian predation
#shouldn't matter really, but this just cuts down unnecesary rows. left_joining creates more rows than df1 if there are duplicate v
#some new rows are created then when joining to the ghost tag df because many of the ghost tags have multiple detections after the
# Im'm just cutting out these excess rows but I think they would get cut down
wghost_av <- left_join(wghost_date, av_pred_df, by = c("TAG")) %>%
    distinct(TAG, Event, Datetime, UTM_X, UTM_Y, first_last, .keep_all = TRUE)
```

The states are then assigned based on above/below the dam, ghost tag, and predation date

The states are then kinda "squished" together into one row on a weekly basis, and duplicated letters in a row are removed, because a fish doesn't need a weekly states history that says "AAAA", it just needs "A". A column is also calculated to get the amount of weekly unique events.

The df then gets rid of duplicated rows, since each fish has multiple rows of states in the same week at this point. Relevant columns are selected and renamed

Some summary information for the fish are then calculated based off these states. In the states_summarized df, the states are "squished" together for all states the fish has. From there, you can see if a fish went above/below the dam, and used the channel, which is what the new columns check. These columns are eventually joined back into the encounter histories summaries wide dataframe.

```
states2 <- states1 %>%
  group_bv(weeks_since, TAG) %>%
  #arranging my datetime ensures that all states will be recorded in the correct order
  arrange(Datetime) %>%
  mutate(
    teststate_2 = paste(state1, collapse = ""),
teststate_3 = gsub('([[:alpha:]])\\1+', '\\1', teststate_2), #removes consecutive letters
    weekly_unique_events = length(unique(Event))
#this is now a weekly chart
states_final <- states2 %>%
 distinct(weeks_since, TAG, teststate_3, .keep_all = TRUE) %>% select(Date, weeks_since, TAG, teststate_3, det_type, ReleaseSite, Species, Release_Length, Release_Weight, c_number_of_detection
  rename(State = teststate_3)
# this makes some columns from all states of fish
states summarized <- states1 %>%
  group by(TAG) %>%
 went_below_dam_throughChannel = str_detect(teststate_3, "BCA"),
went_above_dam_throughChannel = str_detect(teststate_3, "ACB"),
         entered_channel_from_DS = str_detect(teststate_3,
         entered_channel_from_US = str_detect(teststate_3, "BC"),
  select(TAG, went_above_dam_noChannel, went_below_dam_noChannel,went_below_dam_throughChannel,went_above_dam_throughChannel,entere
  distinct(TAG, .keep_all = TRUE)
```

This is an idea to put all the wonky states in place so it would be easy to see, but it isn't fully compatible with the app yet and therefore isn't used. My idea is that instead they would be in one tab on the states page so less one manual parsing is needed.

Shapefile/Polygon Readins

The file "map_polygon_readins.R" reads in all shapefiles for the movements map. Also contains some graphics options for release sites, stationary antenna sites, and labels for stations.

```
#mapping
source("map_polygon_readins.R")|
```

.Rds files

.rds files in R are a binary format used to save R objects, such as data frames, lists, or models, preserving their internal structure and attributes. Unlike CSV files, which store data in a plain text format, .rds files maintain the integrity of R-specific data types and structures, allowing for more complex objects to be saved and loaded efficiently. For example, if you have a data frame df that you want to save as an .rds file, you can use the saveRDS(df, "data.rds") function. To load this data back into R, you would use df <- readRDS("data.rds"). Editing an .rds file involves loading the object, making the desired changes, and then saving it again.

Advantages of .rds files in this app

- Easy to save data and reload it again across different R projects without losing any metadata or attributes
 - o Helpful when saving Stationary data from CombineDataApp and moving over to this app
- Preserves data types and structures
 - Ensures that when the data is read into R, it preserves the same format, vs a csv which saves data in plaintext, sometimes necessitating type conversion to change the data back to its original structure when opened in Excel. This is currently seen when we open and save data in Excel that have tag numbers, and every time we save/modify it in Excel, we have to make sure the tag number is saved as a numeric type rather than general (see Creating New Flat Files/Running the Script: Common errors). This was also causing issues with the timestamps in the stationary data, leading to the decision to keep the Stationary Data in .rds form.
- Ability to save objects other than DFs
 - This enables us to save lists and spatial files as flat files created in the runscript, reducing computational load within the app for faster speed
- Efficiency
 - .rds files are smaller than csvs of the same data. This leads to quicker read-in times and easier file transfers across projects.

As of late 2023, WGFP_Raw (containing all Stationary Antenna data) has been saved as a .rds file in from the update Data app. The new resultant workflow is as follows as it affects this app.

Raw stationary data output saved from updateData app in .rds format goes in the data directory of the data vis app

Open R project file within the data vis app directory

Navigate to, open, and Run or Knit CreateFlatFiles Runscript.rmd Navigate to, open, and run Data Visualization app

- The resulting <u>dataframe</u> is no different than the <u>dataframe</u> output from the previous process
- Biomark process is unaffected
- Current Stationary file size of .rds file vs .csv: ~13 mb vs ~157 mb
- Creates static files used in the data visualization app by cleaning and combining with other files in "data" directory
- Uses the same files as before in the "data" directory of the data visualization app except the stationary file is a .rds file
- Outputs .rds files into the flatFilesforApp folder
- Typically takes a little less than 5 minutes
- Needs to be run anytime there is a change to a data file to get the most up to date data in the app; this includes files in the "data" directory and updated spatial .shp files in the "gis" directory
- App loading time cut down by at least 3 minutes
- No change to functionality
- Output should be more consistent without the need to clean the files within the app anymore

Shiny Modules

Each tab is split up into its own module. This makes the code reusable (can plug the ui and server into any app easily with the right arguments) and organized. This is where the data processing occurs from the sidebar filters and how graphs and tables are rendered. Read more about Shiny Modules here.

Site, Movement and Species Colors

The colors of the Sites, movements and species used in the app are determined by this chunk of code in the beginning of the app.r file

```
#colors
#rainbow trout color pallete used to assign to Sites:
# "Confluence, Connectivity Downstream, Connectivity Side Channel, Connectivity Upstream, Hitching Post,
# Kaibab Park, Red Barn, Windy Gap Auxiliary, Windy Gap Bypass Channel"|
rainbow_trout_colors <- c("#8B8000", "#008080", "#FF6984", "#FF4500", "#6A5ACD", "#32CD32", "#20B2AA", "#FF1C55", "#4682B4
#currently "Changed Rivers", "Downstream Movement" "Initial Release", "No Movement", "Upstream Movement" (5/17/24)
#note: these are a little different than what we see in the map because the map/marker color optoins are very limited.
movementColors <- c("#4B0082", "#8B0000", "#FF8C00", "#253333", "#22bd74") #, "#66FF00"
# currently "LOC" "MTS" "RBT" "RXN" "TGM" (5/17/24)
speciesColors <- c("#FFD700", "#654321", "#4F7942", "#FF7F50", "#1E90FF", "#008080", "#DAA520", "#DAA520", "#9A5ECD")</pre>
```

Rainbow_trout_colors is used to assign colors to site in succession of the colors: so Confluence gets assigned #8b8000, Connectivity Downstream is assigned #008080, etc. When a new site is added to the data, it will automatically get assigned a color based off the rainbow_trout_colors vector, provided that the number of sites is less than the length of rainbow_trout_colors. As of 5/24/2024, there are 9 sites and 15 colors in the vector, so we can add 6 more antenna sites that won't require adjustment to this code.

MovementColors and SpeciesColors are similar to rainbow_trout_colors. If we want to change the colors for a given species/movement etc, then change the colors in this code. Note: changing movementColors does not affect the color shown in the map, just the graphs. Go to get_movements_function.R to edit the movement colors on the map.

Let me know if there are other questions and concerns, I'm happy to add to this if needed. Let me know sfigraf@gmail.com