

# R Training Documentation

2023-10-17

## Introduction

### Source files, directories, important functions in R

The working directory is the folder your R session ‘lives in.’ In general, it is good practice to have your working directory to be set to the folder which houses the open file. `getwd()` displays the current working directory of the session. You can set the working directory to the path of the desired folder by using `setwd(...)`.

If ever you are confused about a package or function in R studio, typing ‘?’ in front of the argument will display the documentation for that function in your session. For information on Vectorsurv specific functions, please see the documentation below.

```
#To get current directory  
getwd()
```

```
## [1] "C:/Users/Christina/Desktop/R-tutorial"
```

```
#To get information on function in R documentation  
?getwd()  
  
#'VS_functions.R' should be in the same directory as the .rmd  
#'files which access its functions, you can set the working directory  
  
#Path to folder on my computer: "C:\Users\Christina\Desktop\R-tutorial"  
#this path will be slightly different for you depending on the file location.  
  
#Set below to the location of the downloaded files on your machine,  
#make sure to change the direction of the slash from "\" to "/" to indicate opening of the folder.  
  
setwd("/Users/Christina/Desktop/R-tutorial")
```

There is one R files. The .R file is a source file that contains all of the functions you will need to run the .RMarkdown files. You do not need to touch the source file. The .RMarkdown is a pre-coded sample report that you can customize to suit your needs.

```
#loads relevant packages and functions  
source("VS_functions.R")
```

## Retriving Data

### **getArthroCollections(...)**

#### *Description*

getArthroCollections(...) obtains collections data on a year range (start\_year, end\_year) and agency\_id. It prompts the user for their Gateway username and password before retrieving the associated data. Agency id is the number associated with your agency through Vectorsurv. If you have access to multiple agencies, id can be used to specific what data you wish to retrieve. You can only retrieve data from agencies linked to your Gateway account.

#### *Usage*

```
getArthroCollections(start_year, end_year, agency_id)
```

#### *Arguments*

- start\_year: Beginning of year range
- end\_year: End of year range
- agency\_id: Agency ID number, if left blank, the default will return data for all agencies linked to you account. The majority of users are linked to one agency

#### *#Example*

```
collections = getArthroCollections(2022,2023, 55)
```

### **getPools(...)**

#### *Description*

getPools(...) similar to getArthroCollections() obtains pools on a year range (start\_year, end\_year) and agency\_id. It prompts the user for their Gateway username and password before retrieving the associated data. getPools() retrieve data for both mosquito and tick pools.

#### *Usage*

```
getPools(start_year, end_year, agency_id)
```

#### *Arguments*

- start\_year: Beginning of year range
- end\_year: End of year range
- agency\_id: Agency ID number

#### *#Example*

```
pools = getPools(2022,2023, 55)
```

## Write Data to file

You can save retrieved data as a .csv file in your current directory using write.csv(). That same data can be retrieved using read.csv(). Writing data to a .csv can make the rendering process more efficient when generating reports in R. We recommend that you write the data pulled from our API into a csv and then load that data when generating reports.

```
read.csv(...)
```

```
#creates a file named "collections_18_23.csv" in your current directory
write.csv(x = collections, file = "collections_18_23.csv")

#loads collections data
collections = read.csv("collections_18_23.csv")
```

## Basic subsetting, filtering, grouping, pivoting

Data can be subset to contain columns of interest. Sub-setting can also be used to reorder the columns in a dataframe. Do not subset collections or pools data before inputting them into Vectorsurv calculator functions to avoid losing essential columns. It is recommended to subset after calculations are complete and before inputting into a table generator. **Remember, subsetting, filtering, grouping and summarizing will not change the value of the data unless it is reassigned to the same variable name.** We recommend creating a new variable for processed data.

### Subsetting

```
#Subset using column names or index number

colnames(collections) #displays column names and associated index
```

```
## [1] "X" "collection_id"
## [3] "collection_num" "collection_date"
## [5] "collection_date_date_only" "comments"
## [7] "identified_by" "num_trap"
## [9] "site" "surv_year"
## [11] "trap_nights" "trap_problem_bit"
## [13] "user" "add_date"
## [15] "deactive_date" "updated"
## [17] "id" "num_count"
## [19] "sex_id" "sex_type"
## [21] "sex_name" "species_id"
## [23] "species_full_name" "species_display_name"
## [25] "agency_id" "agency_code"
## [27] "agency_name" "trap_id"
## [29] "trap_acronym" "trap_name"
## [31] "trap_presence"
```

```
#Subsetting by name
head(collections[c("collection_date", "species_display_name", "num_count")])
```

```
##      collection_date species_display_name num_count
## 1 2023-10-24T07:00:00.000Z      Cs inornata         1
## 2 2023-10-24T07:00:00.000Z      Cx pipiens        11
## 3 2023-10-24T07:00:00.000Z      Cx tarsalis         1
## 4 2023-10-24T07:00:00.000Z      Ae aegypti         1
## 5 2023-10-24T07:00:00.000Z      Ae melanimon         1
## 6 2023-10-24T07:00:00.000Z      Cx pipiens        13
```

```
#by index
head(collections[c(3,23,17)])
```

```
##   collection_num species_full_name      id
## 1           8444 Culiseta inornata 6232791
## 2           8444      Culex pipiens 6232789
## 3           8444      Culex tarsalis 6232790
## 4           8444      Aedes aegypti 6232788
## 5           8445      Aedes melanimon 6232808
## 6           8445      Culex pipiens 6232809
```

```
#to save a subset
collections_subset = collections[c(3,23,17)]
```

## Filtering and subsetting in dplyr

Dplyr is a powerful package for filtering and sub-setting data. It follows logic similar to SQL queries.

For more information on data manipulation using dplyr [Click Here](#)

Dplyr utilizes the pipe operator ‘%>%’ to send data into functions.

```
#Subsetting columns with dplyr 'select'
collections %>%
  select(collection_date,species_display_name,num_count)%>%
  head(1)
```

```
##           collection_date species_display_name num_count
## 1 2023-10-24T07:00:00.000Z          Cs inornata          1
```

```
#filtering with dplyr 'filter'
collections %>%
  filter(species_display_name=="Cx pipiens") %>%
  head(1)
```

```
##   X collection_id collection_num      collection_date
## 1 2           3035586           8444 2023-10-24T07:00:00.000Z
##   collection_date_date_only
## 1                TRUE
##
## 1 Week 7 Trap Comparison Project\nWith Lure\nCatch Bag: F1\n\t\t\t\tSubmitted using API - MVTrapID# 8
##   identified_by num_trap  site surv_year trap_nights trap_problem_bit user
## 1 Whitney;Clack      1 119819    2023          1          FALSE 331
##           add_date deactivate_date          updated      id
## 1 2023-10-24T19:14:09.015Z          NA 2023-10-24T19:14:09.519Z 6232789
##   num_count sex_id sex_type      sex_name species_id species_full_name
## 1         11     4  female Females - Mixed          65      Culex pipiens
##   species_display_name agency_id agency_code          agency_name trap_id
## 1          Cx pipiens      55      SAYO Sacramento-Yolo MVCD          13
##   trap_acronym  trap_name trap_presence
## 1      BGSENT BG Sentinel          FALSE
```

*#filtering multiple arguments using '%in%'*

```
collections %>%
  filter(species_display_name %in% c("Cx pipiens", "Cx tarsalis"),
         num_trap %in% c(1:5))%>%
  head(1)
```

```
##   X collection_id collection_num      collection_date
## 1 2          3035586          8444 2023-10-24T07:00:00.000Z
##   collection_date_date_only
## 1                        TRUE
##
## 1 Week 7 Trap Comparison Project\nWith Lure\nCatch Bag: F1\n\t\t\t\tSubmitted using API - MVTrapID# 8
##   identified_by num_trap  site surv_year trap_nights trap_problem_bit user
## 1 Whitney;Clack      1 119819    2023          1          FALSE 331
##           add_date deactivate_date          updated      id
## 1 2023-10-24T19:14:09.015Z          NA 2023-10-24T19:14:09.519Z 6232789
##   num_count sex_id sex_type      sex_name species_id species_full_name
## 1         11     4  female Females - Mixed          65      Culex pipiens
##   species_display_name agency_id agency_code      agency_name trap_id
## 1          Cx pipiens      55      SAYO Sacramento-Yolo MVCD      13
##   trap_acronym  trap_name trap_presence
## 1      BGSENT BG Sentinel          FALSE
```

## Group by

In addition to filtering and sub-setting, data can be group by variables and summarized.

*#groups by species and collection date and sums the number counted*

```
collections %>%
  group_by(collection_date, species_display_name) %>%
  summarise(sum_count = sum(num_count, na.rm=T))%>%
  head()
```

```
## # A tibble: 6 x 3
## # Groups:   collection_date [1]
##   collection_date      species_display_name sum_count
##   <chr>              <chr>                <int>
## 1 2022-01-04T08:00:00.000Z An freeborni             1
## 2 2022-01-04T08:00:00.000Z Cs incidens              2
## 3 2022-01-04T08:00:00.000Z Cs inornata               2
## 4 2022-01-04T08:00:00.000Z Cx pipiens                22
## 5 2022-01-04T08:00:00.000Z Cx stigmatosoma            2
## 6 2022-01-04T08:00:00.000Z Cx tarsalis                1
```

*#groups by species and collection date and takes the average the number counted*

```
collections %>%
  group_by(collection_date, species_display_name) %>%
  summarise(avg_count = mean(num_count, na.rm=T))%>%
  head()
```

```
## # A tibble: 6 x 3
## # Groups:   collection_date [1]
##   collection_date      species_display_name avg_count
##   <chr>                <chr>                <dbl>
## 1 2022-01-04T08:00:00.000Z An freeborni          1
## 2 2022-01-04T08:00:00.000Z Cs incidens           1
## 3 2022-01-04T08:00:00.000Z Cs inornata           1
## 4 2022-01-04T08:00:00.000Z Cx pipiens           2.44
## 5 2022-01-04T08:00:00.000Z Cx stigmatosoma       1
## 6 2022-01-04T08:00:00.000Z Cx tarsalis           1
```

## Pivoting

Data can be manipulated into long and wide (spread sheet) forms using `pivot_wider` and `pivot_longer`. By default data from the API is in long form. Here we pivot on species and sex condition names using `num_count` as values. The end result is data with `num_count` values in the columns named `species_sex`. For more on pivoting see `??pivot_longer` and `??pivot_wider`

```
collections_wide = pivot_wider(collections,
                               names_from = c("species_display_name", "sex_name"),
                               values_from = "num_count")
```

## Calculations

### Abundance

#### `getAbundance(...)`

##### *Description*

`getAbundance(...)` uses any amount of arthro collections data to calculate the abundance for the specified parameters. The function calculates using the methods of the Gateway Abundance calculator.

##### *Usage*

```
getAbundance(collections, interval, species_list = NULL, trap_list = NULL, species_seperate = FALSE)
```

##### *Arguments*

- `collections`: Collections data retrieved from `getArthroCollections(...)`
- `interval`: Calculation interval for abundance, accepts “collection\_date”, “Biweek”, “Week”, and “Month.”
- `species_list`: Species filter for calculating abundance. `Species_display_name` is the accepted notation. To see a list of species present in your data run `unique(collections$species_display_name)`. If species is unspecified, the default NULL will return data for all species in data.
- `trap_list`: Trap filter for calculating abundance. `Trap_acronym` is the accepted notation. Run `unique(collections$trap_acronym)` to see trap types present in your data. If `trap_list` is unspecified, the default NULL will return data for all trap types.
- `species_seperate`: Should the species in `species_list` have abundance calculated separately? Setting to FALSE calculates the combined abundance. The same result can be performed by calculating on one species at the time.

```
getAbundance(collections,
             interval = "Biweek",
             species_list = c("Cx tarsalis", "Cx pipiens"),
             trap_list = "C02",
             species_separate = FALSE)
```

##	EPIYEAR	Biweek	Count	Trap_Events	Abundance
## 1	2023	10	882	65	13.57
## 2	2023	11	3254	142	22.92
## 3	2023	12	4395	153	28.73
## 4	2023	13	15803	182	86.83
## 5	2023	14	24939	226	110.35
## 6	2023	15	24113	217	111.12
## 7	2023	16	19062	255	74.75
## 8	2023	17	12865	226	56.92
## 9	2023	18	10088	213	47.36
## 10	2023	19	7161	211	33.94
## 11	2023	20	5934	211	28.12
## 12	2023	21	2806	144	19.49
## 13	2023	22	16	4	4.00
## 14	2022	4	9	3	3.00
## 15	2022	9	1358	126	10.78
## 16	2022	10	1202	133	9.04
## 17	2022	11	1969	145	13.58
## 18	2022	12	3503	159	22.03
## 19	2022	13	5630	159	35.41
## 20	2022	14	10444	154	67.82
## 21	2022	15	9722	178	54.62
## 22	2022	16	7949	186	42.74
## 23	2022	17	6501	180	36.12
## 24	2022	18	6038	166	36.37
## 25	2022	19	3798	163	23.30
## 26	2022	20	1869	120	15.57
## 27	2022	21	1189	84	14.15

## Abundance Anomaly (comparison to 5 year average)

### getAbundanceAnomaly()

#### Description

getAbundanceAnomaly(..) requires at least five years prior to the target\_year of arthro collections data to calculate for the specified parameters. The function uses the methods of the Gateway Abundance Anomaly calculator, and will not work if there is fewer than five years of data present.

#### Usage

```
getAbundanceAnomaly(collections,interval,target_year, species_list = NULL, trap_list = NULL,
species_separate = FALSE)
```

#### Arguments

- collections: Collections data retrieved from getArthroCollections(..)
- interval: Calculation interval for abundance, accepts “collection\_date”, “Biweek”, “Week”, and “Month”.

- `target_year`: Year to calculate analysis on. Collections data must have a year range of at least (`target_year - 5`, `target_year`).
- `species_list`: Species filter for calculating abundance. `Species_display_name` is the accepted notation. To see a list of species present in your data run `unique(collections$species_display_name)`. If species is unspecified, the default NULL will return data for all species in data.
- `trap_list`: Trap filter for calculating abundance. `Trap_acronym` is the accepted notation. Run `unique(collections$trap_acronym)` to see trap types present in your data. If `trap_list` is unspecified, the default NULL will return data for all trap types.
- `species_separate`: Should the species in `species_list` have abundance calculated separately? Setting to FALSE calculates the combined abundance. The same result can be performed by calculating on one species at the time.

```
collections_18_23 = getArthroCollections(2018,2023, 55)

getAbundanceAnomaly(collections_18_23,
  interval = "Biweek",
  target_year = 2023,
  species_list = c("Cx tarsalis", "Cx pipiens"),
  trap_list = "C02",
  species_separate = FALSE)
```

##	Biweek	EPIYEAR	Count	Trap_Events	Abundance	Five_Year_Avg	Delta
## 1	10	2023	882	65	13.57	12.926	4.98
## 2	11	2023	3254	142	22.92	19.666	16.55
## 3	12	2023	4395	153	28.73	37.988	-24.37
## 4	13	2023	15803	182	86.83	54.496	59.33
## 5	14	2023	24939	226	110.35	81.972	34.62
## 6	15	2023	24113	217	111.12	75.588	47.01
## 7	16	2023	19062	255	74.75	78.528	-4.81
## 8	17	2023	12865	226	56.92	66.406	-14.28
## 9	18	2023	10088	213	47.36	61.704	-23.25
## 10	19	2023	7161	211	33.94	51.736	-34.40
## 11	20	2023	5934	211	28.12	32.970	-14.71
## 12	21	2023	2806	144	19.49	20.082	-2.95
## 13	22	2023	16	4	4.00	18.330	-78.18

## Infection Rate

### getInfectionRate()

#### Description

`getInfectionRate(..)` requires at least five years prior to the `target_year` of arthro collections data to calculate for the specified parameters. The function uses the methods of the Gateway Abundance Anomaly calculator, and will not work if there is fewer than five years of data present.

#### Usage

```
getInfectionRate(pools,interval, target_year, target_disease,pt_estimate, species_list = c(NULL), trap_list = c(NULL))
```

#### Arguments

- `pools`: Pools data retrieved from `getPools(...)`
- `interval`: Calculation interval for abundance, accepts “collection\_date”, “Biweek”, “Week”, and “Month”.



- `target_year`: Year to calculate infection rate for. This year must be present in the data.
- `target_disease`: The disease to calculate infection rate for—i.e. “WNV”. Disease acronyms are the accepted input. To see a list of disease acronyms, run `unique(pools$target_acronym)`.
- `pt_estimate`: The estimation type for infection rate. Options include: “mle”, “bc-”mle”, “mir.”
- `species_list`: Species filter for calculating abundance. `Species_display_name` is the accepted notation. To see a list of species present in your data run `unique(pools$species_display_name)`. If species is unspecified, the default NULL will return data for all species in data.
- `trap_list`: Trap filter for calculating abundance. `Trap_acronym` is the accepted notation. Run `unique(pools$trap_acronym)` to see trap types present in your data. If `trap_list` is unspecified, the default NULL will return data for all trap types.

```
IR = getInfectionRate(pools,
                      interval = "Week",
                      target_year = 2023,
                      target_disease = "WNV",
                      pt_estimate = "mle",
                      species_list = c("Cx pipiens"),
                      trap_list = c("CO2", "GRVD") )
IR
```

##	Year	Week	Disease	Point_Estimate	Lower_CI	Upper_CI
## 1	2023	20	WNV	0.0000000	0.00000000	4.617179
## 2	2023	21	WNV	0.0000000	0.00000000	4.119261
## 3	2023	22	WNV	0.0000000	0.00000000	3.156551
## 4	2023	23	WNV	0.5727378	0.03289081	2.738134
## 5	2023	24	WNV	0.0000000	0.00000000	1.851039
## 6	2023	25	WNV	0.5137433	0.02945432	2.465164
## 7	2023	26	WNV	3.6102945	2.00908328	5.993620
## 8	2023	27	WNV	5.7702646	3.12648843	9.759822
## 9	2023	28	WNV	5.3827182	3.25546076	8.369922
## 10	2023	29	WNV	9.9744640	6.82440531	14.033134
## 11	2023	30	WNV	10.7927309	6.23631244	17.371132
## 12	2023	31	WNV	10.5232311	6.77649116	15.512533
## 13	2023	32	WNV	11.7049730	6.70172960	18.933345
## 14	2023	33	WNV	11.1300732	6.54666228	17.597019
## 15	2023	34	WNV	17.6982563	11.78626386	25.389190
## 16	2023	35	WNV	11.1131842	5.69748179	19.473935
## 17	2023	36	WNV	8.1580616	4.17956384	14.365812
## 18	2023	37	WNV	3.9879779	1.63947663	8.160978
## 19	2023	38	WNV	2.7297894	0.72200878	7.245097
## 20	2023	39	WNV	0.0000000	0.00000000	2.223982
## 21	2023	40	WNV	1.8502043	0.33078143	5.950046
## 22	2023	41	WNV	1.0256410	0.05906777	4.851477
## 23	2023	42	WNV	0.0000000	0.00000000	11.530964

## Vector Index

### getVectorIndex()

#### Description

`getVectorIndex()(..)` requires at least five years prior to the `target_year` of arthro collections data to calculate for the specified parameters. The function uses the methods of the Gateway Abundance Anomaly calculator, and will not work if there is fewer than five years of data present.

## Usage

`getVectorIndex(collections, pools, interval, target_year, target_disease, pt_estimate, species_list=NULL, trap_list = NULL)`

*Arguments* - `collections`: collections data retrieved from `getCollections(...)` - `pools`: Pools data retrieved from `getPools(...)`

**Note: Years from pools and collections data must match**

- `interval`: Calculation interval for abundance, accepts “collection\_date”, “Biweek”, “Week”, and “Month.”
- `target_year`: Year to calculate infection rate for. This year must be present in the data.
- `target_disease`: The disease to calculate infection rate. Disease acronyms are the accepted input. To see a list of disease acronyms, run `unique(pools$target_acronym)`.
- `pt_estimate`: The estimation type for infection rate. Options include: “mle”, “bc-mle”, “mir.”
- `species_list`: Species filter for calculating abundance. `Species_display_name` is the accepted notation. To see a list of species present in your data run `unique(pools$species_display_name)`. If species is unspecified, the default NULL will return data for all species in data.
- `trap_list`: Trap filter for calculating abundance. `Trap_acronym` is the accepted notation. Run `unique(pools$trap_acronym)` to see trap types present in your data. If `trap_list` is unspecified, the default NULL will return data for all trap types.

```
pools=getPools(2023,2023,55)
collections= getArthroCollections(2023,2023,55)

getVectorIndex(collections, pools, interval = "Biweek",
               2023,
               target_disease = "WNV", pt_estimate = "bc-mle",
               species_list=c("Cx tarsalis"),

               trap_list = c("CO2"))
```

##	Biweek	EPIYEAR	Count	Trap_Events	Abundance	Year	Disease	Point_Estimate
## 1	10	2023	509	65	7.83	2023	WNV	0.0000000
## 2	11	2023	1910	142	13.45	2023	WNV	0.5205304
## 3	12	2023	2343	153	15.31	2023	WNV	0.4359300
## 4	13	2023	12226	182	67.18	2023	WNV	0.9791292
## 5	14	2023	21573	226	95.46	2023	WNV	3.2518724
## 6	15	2023	20979	217	96.68	2023	WNV	7.1369213
## 7	16	2023	17215	255	67.51	2023	WNV	10.3384919
## 8	17	2023	11019	226	48.76	2023	WNV	8.8628801
## 9	18	2023	8184	213	38.42	2023	WNV	5.6198226
## 10	19	2023	4625	211	21.92	2023	WNV	4.6584348
## 11	20	2023	3213	211	15.23	2023	WNV	1.0619013
## 12	21	2023	1486	144	10.32	2023	WNV	0.0000000
##	Lower_CI	Upper_CI	VectorIndex					
## 1	0.00000000	6.735594	0.000000					
## 2	0.02996660	2.523691	7.001133					
## 3	0.02501275	2.117993	6.674089					
## 4	0.43252140	1.931931	65.777899					
## 5	2.37760030	4.355111	310.423742					

```
## 6 5.68541180 8.868693 689.997556
## 7 8.27552335 12.801459 697.951588
## 8 6.94037779 11.190148 432.154034
## 9 3.97673478 7.748289 215.913585
## 10 2.89908373 7.147550 102.112891
## 11 0.27964749 2.876519 16.172757
## 12 0.00000000 3.006583 0.000000
```

## Tables

### getPoolsComparisionTable()

#### Description

getPoolsComparisionTable() produces a frequency table for positive and negative pools counts by year and species. The more years present in the data, the larger the table.

#### Usage

```
getPoolsComparisionTable(pools,target_disease, species_seperate=F)
```

#### Arguments

- pools: Pools data retrieved from getPools(...)
- target\_disease: The disease to calculate infection rate for-i.e. “WNV”. Disease acronyms are the accepted input. To see a list of disease acronyms, run unique(pools\$target\_acronym).
- species\_seperate: Should the pools comparison be split by species of each pool. Default is FALSE.

```
getPoolsComparisionTable(pools, "WNV", species_seperate = T)
```

```
## # A tibble: 3 x 6
## # Groups:   surv_year, species_display_name [3]
##   surv_year species_display_name Negative Confirmed Total PercentPositive
##   <int> <chr> <int> <int> <int> <dbl>
## 1 2023 An freeborni 1 0 1 0
## 2 2023 Cx pipiens 3833 243 4076 5.96
## 3 2023 Cx tarsalis 3413 388 3801 10.2
```

## Styling Dataframes with kable

Professional looking tables can be produced using the kable and kableExtra packages.

```
AbAnOutput = getAbundance(collections,
                          interval = "Biweek",

                          species_list = c("Cx tarsalis", "Cx pipiens"),
                          trap_list = "C02",
                          species_seperate = FALSE)
head(AbAnOutput)
```

```
##   EPIYEAR Biweek Count Trap_Events Abundance
## 1 2023 10 882 65 13.57
```

Disease Year	Biweek	Count	Trap Events	Abundance
2023	10	882	65	13.57
2023	11	3254	142	22.92
2023	12	4395	153	28.73
2023	13	15803	182	86.83
2023	14	24939	226	110.35
2023	15	24113	217	111.12
2023	16	19062	255	74.75
2023	17	12865	226	56.92
2023	18	10088	213	47.36
2023	19	7161	211	33.94
2023	20	5934	211	28.12
2023	21	2806	144	19.49
2023	22	16	4	4.00

Table X: Combined biweekly Abundance Calculation for Cx. tarsalis, pipiens in CO2 traps

```
## 2    2023    11  3254      142    22.92
## 3    2023    12  4395      153    28.73
## 4    2023    13 15803      182    86.83
## 5    2023    14 24939      226   110.35
## 6    2023    15 24113      217   111.12
```

*#As a kable table where column names, font\_size, type and much more can be customized*

```
AbAnOutput %>%
  kbl(col.names = c("Disease Year", "Biweek", "Count", "Trap Events", "Abundance")) %>%
  kable_styling(bootstrap_options = "striped",
                font_size = 14,
                latex_options="scale_down") %>%
  footnote(general = "Table X: Combined biweekly Abundance Calculation for Cx. tarsalis, pipiens in CO2
```

## Data using datatables

Interactive html only tables can be produced using the DT package. DT tables allow for sorting and filtering with in a webpage. These are ideal for viewing data but are not compatible with pdf or word formats.

```
#AbAnOutput %>%
  #datatable(colnames = c("Disease Year", "Biweek", "Count", "Trap Events", "Abundance"))
```

## Charts and Graphs

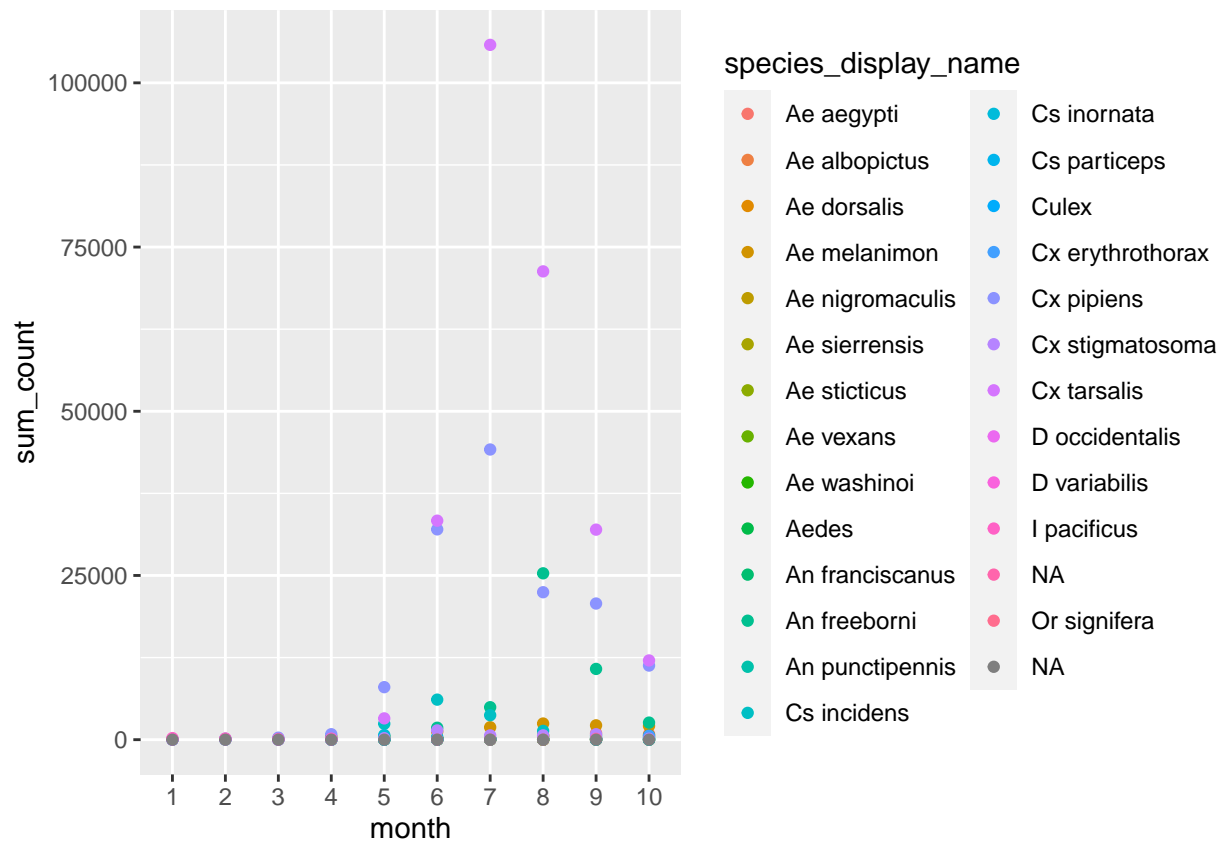
Ggplot is a easy to use plotting library in R. Gplot syntax consists of creating a ggplot object with a dataframe and adding subsequent arguments to that object. Aesthetics (aes) in ggplot represents the data mapping aspect of the plot. A simple example using collections is shown below.

```
#creates a month column and translates numerics
collections$month = as.factor(month(collections$collection_date))
```

```
collections_sums = collections %>%
  group_by(month, species_display_name) %>%
  summarise(sum_count = sum(num_count, na.rm=T))
```

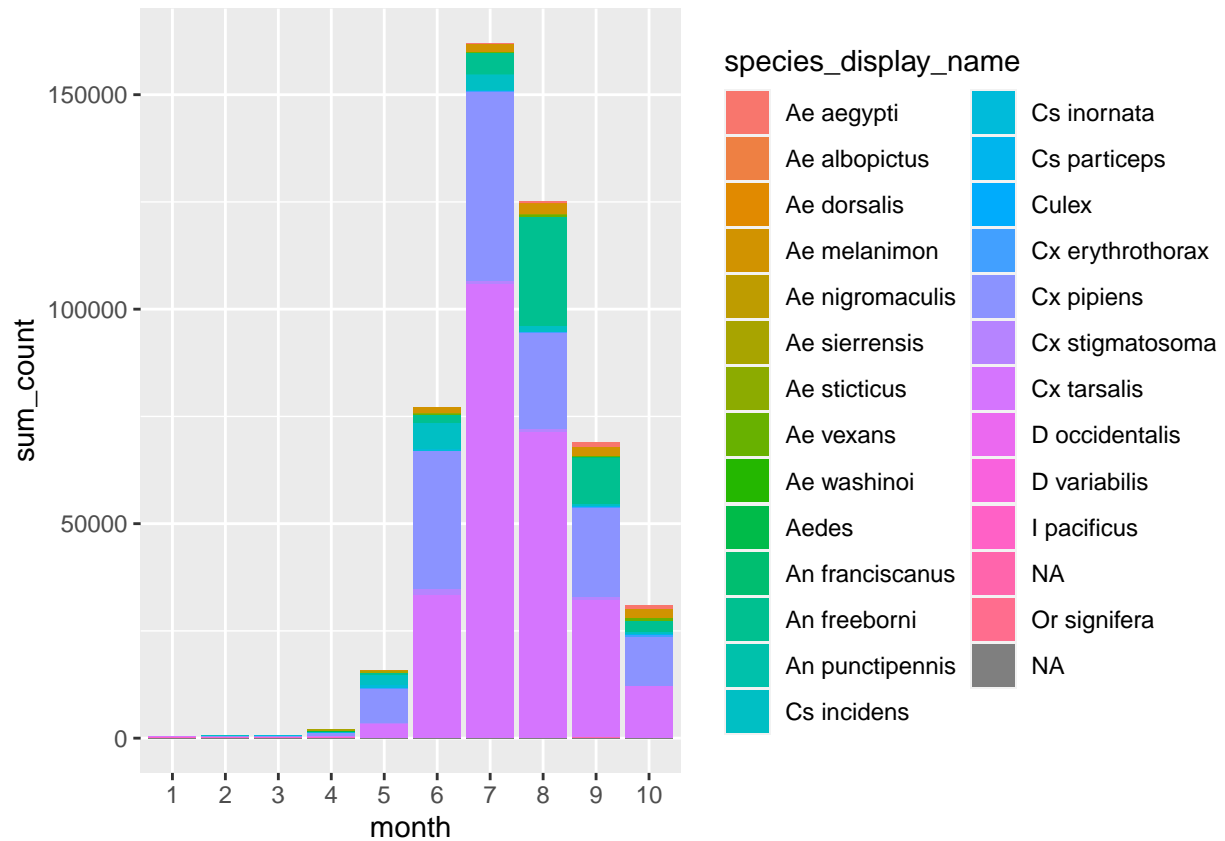
*#ggplot with dots a values for each species*

```
ggplot(data=collections_sums,
  aes(x = month, y = sum_count, color = species_display_name))+
  geom_point()
```

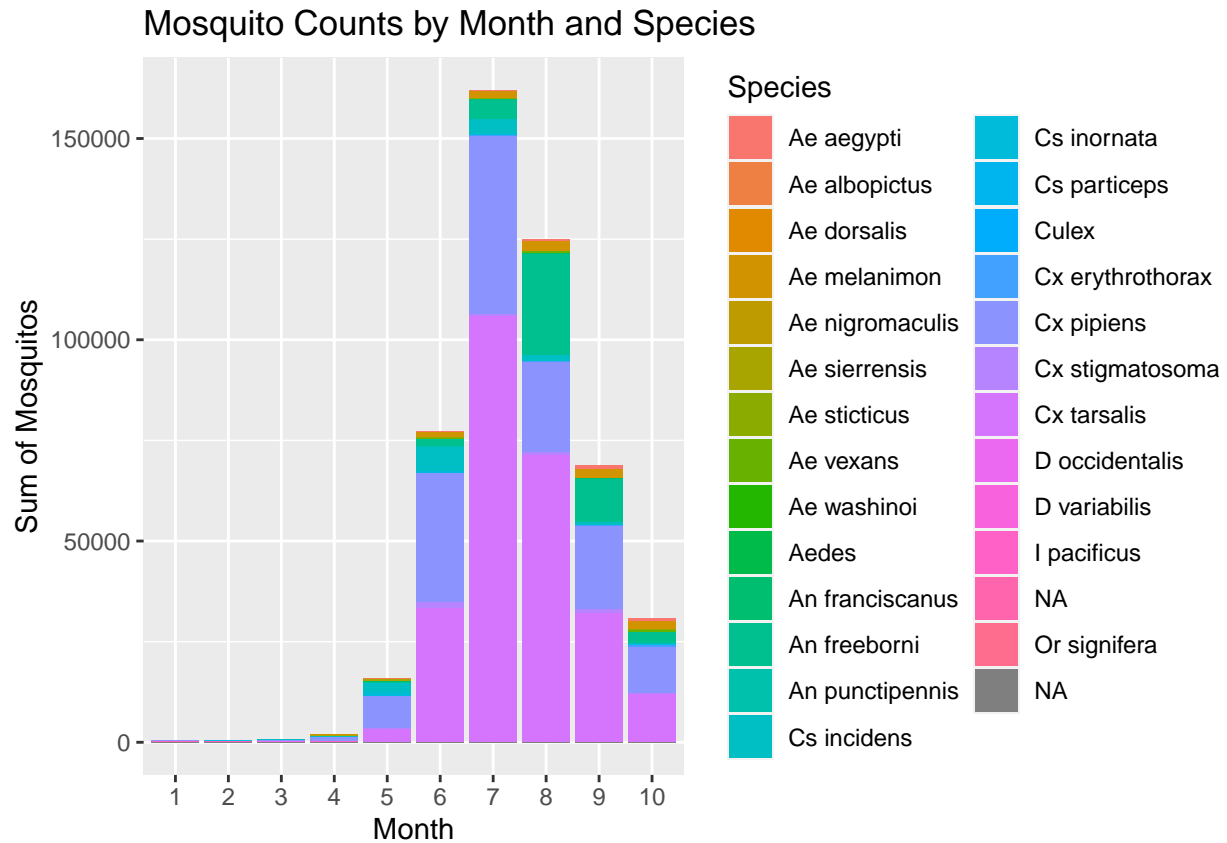


*#bar chart*

```
ggplot(data=collections_sums,
  aes(x = month, y = sum_count, fill = species_display_name))+
  geom_bar(stat="identity")
```



```
#adding labels
ggplot(data=collections_sums,
  aes(x = month, y = sum_count, fill = species_display_name))+
  geom_bar(stat="identity") +
  labs(title = "Mosquito Counts by Month and Species",x = "Month",
    y = "Sum of Mosquitos",
    fill = "Species")
```



When plotting with libraries in R, it is easiest when the data is prepared in long form. Most calculator outputs from our functions are in wide form. The following wrapper functions help process and plot this data.

#### ProcessAbunAnom()

##### Description

ProcessAbunAnom() processes the output returned from getAbundanceAnomaly() into a long form suitable for plotting in ggplot.

##### Usage

```
ProcessAbunAnom(AbAnomOutput)
```

##### Arguments

- AbAnomOutput: Output from returned getAbundanceAnomaly()

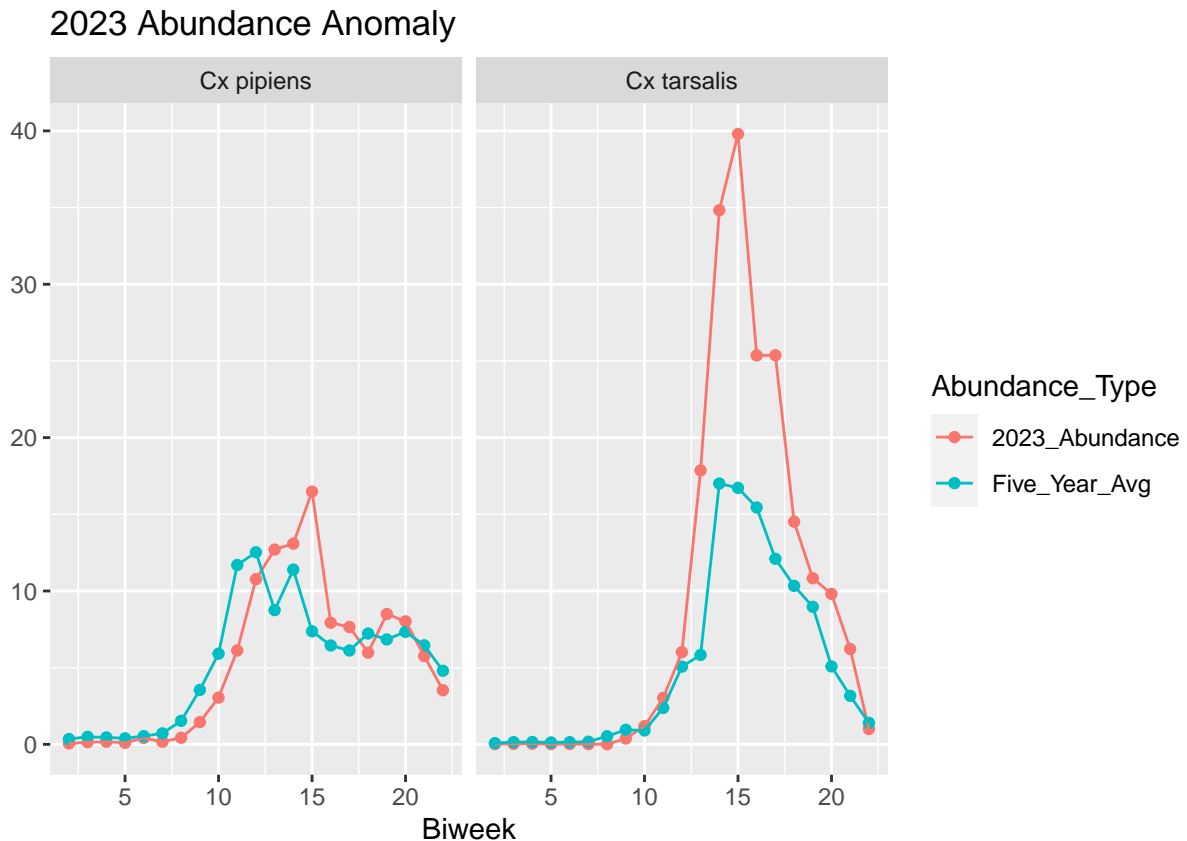
```
AbAnOut = getAbundanceAnomaly(collections_18_23,
                              interval = "Biweek",
                              target_year = 2023,
                              species_list = c("Cx tarsalis", "Cx pipiens"),
                              species_seperate = TRUE) #species_seperate set to true will allow for us

AbAnOut_L = ProcessAbunAnom(AbAnOut)

#Example using ggplot
```

```
AbAnOut_L %>% filter(Abundance_Type %in% c("2023_Abundance",
                                             "Five_Year_Avg"))%>%

  ggplot(aes(x=Biweek,
             y= Abundance_Calculation,
             color = Abundance_Type)) +
  geom_point()+
  geom_line() +
  facet_wrap(~species_display_name)+
  labs(title="2023 Abundance Anomaly",y = "")
```



## plotInfectionRate()

### Description

plotInfectionRate() plots the output returned from getInfectionRate() with confidence intervals using ggplot

### Usage

```
plotInfectionRate(InfRtOutput)
```

### Arguments

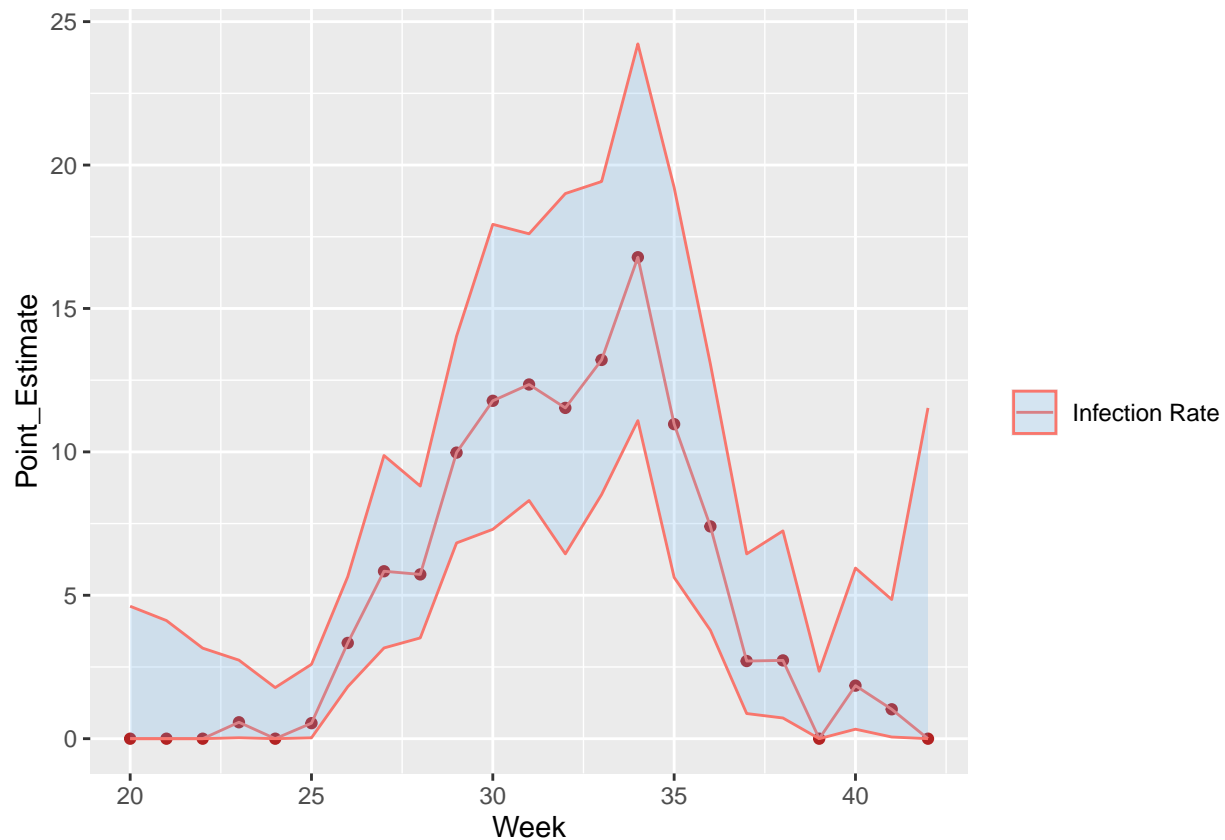
- InfRtOutput: Output from returned getInfectionRate()

```
IR = getInfectionRate(pools,
                      interval = "Week",
                      target_year = 2023,
```



```
target_disease = "WNV",
pt_estimate = "mle",
species_list = c("Cx pipiens"),
trap_list = c("CO2","GRVD") )

plotInfectionRate(InfRtOutput = IR)
```



## Additional Charting Examples

We can highlight rows and columns, add headers, and customize footnotes. For more information please [Click Here](#)

```
collections= getArthroCollections(2021,2023,55)

table(collections$trap_acronym,collections$surv_year)%>%
  kbl(align = "c") %>%
  kable_paper(full_width = F,
    html_font="arial",
    lightable_options = "striped",
  ) %>%
  add_header_above(c("Trap Type", "Years" = 3)) %>%
  footnote(general = "Table X: Traps deployed by year", general_title = "") %>%
  row_spec(c(3,9,10), background = "yellow")%>%
  column_spec(c(4), background = "orange")
```

Trap Type	Years		
	2021	2022	2023
BACKPACK	26	33	11
BGSENT	5600	7139	6456
BTLJC	84	0	0
CO2	6218	5488	7063
FLANNEL	301	296	172
GRVD	8270	7700	7364
LCKR	3707	3693	2787
OTHER	10	11	36
OVI	0	294	0
WRKR	124	0	0

Table X: Traps deployed by year