National Parks and Biodiversity - Demographics Report

Name: Veronica Alejandro

UT EID: vaa678

Homework 5

This is our last homework. We will combine some of the concepts from our recent lab on visualizing geospatial data with merging data and create an interactive dashboard.

In [1]: !pip3 install vega-datasets

Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: vega-datasets in /Users/vero/Library/Python/3.9/lib/python/site-packages (0.9.0)

Requirement already satisfied: pandas in /Users/vero/Library/Python/3.9/lib/python/site-packages (from veg a-datasets) (1.5.3)

Requirement already satisfied: pytz>=2020.1 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from pandas->vega-datasets) (2022.7.1)

Requirement already satisfied: python-dateutil>=2.8.1 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from pandas->vega-datasets) (2.8.2)

Requirement already satisfied: numpy>=1.20.3 in /Users/vero/Library/Python/3.9/lib/python/site-packages (f rom pandas->vega-datasets) (1.24.1)

Requirement already satisfied: six>=1.5 in /Library/Developer/CommandLineTools/Library/Frameworks/Python3. framework/Versions/3.9/lib/python3.9/site-packages (from python-dateutil>=2.8.1->pandas->vega-datasets) (1.15.0)

[notice] A new release of pip available: 22.3.1 -> 23.1
[notice] To update, run: /Library/Developer/CommandLineTools/usr/bin/python3 -m pip install --upgrade pip

In [2]: !pip3 install altair

```
hwk5
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: altair in /Users/vero/Library/Python/3.9/lib/python/site-packages (4.2.2)
Requirement already satisfied: numpy in /Users/vero/Library/Python/3.9/lib/python/site-packages (from alta
ir) (1.24.1)
Requirement already satisfied: jinja2 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from alt
air) (3.1.2)
Requirement already satisfied: entrypoints in /Users/vero/Library/Python/3.9/lib/python/site-packages (fro
m altair) (0.4)
Requirement already satisfied: toolz in /Users/vero/Library/Python/3.9/lib/python/site-packages (from alta
ir) (0.12.0)
Requirement already satisfied: jsonschema>=3.0 in /Users/vero/Library/Python/3.9/lib/python/site-packages
(from altair) (4.17.3)
Requirement already satisfied: pandas>=0.18 in /Users/vero/Library/Python/3.9/lib/python/site-packages (fr
om altair) (1.5.3)
Requirement already satisfied: attrs>=17.4.0 in /Users/vero/Library/Python/3.9/lib/python/site-packages (f
rom isonschema>=3.0->altair) (22.2.0)
Requirement already satisfied: pyrsistent!=0.17.0,!=0.17.1,!=0.17.2,>=0.14.0 in /Users/vero/Library/Pytho
n/3.9/lib/python/site-packages (from jsonschema>=3.0->altair) (0.19.3)
Requirement already satisfied: python-dateutil>=2.8.1 in /Users/vero/Library/Python/3.9/lib/python/site-pa
ckages (from pandas>=0.18->altair) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /Users/vero/Library/Python/3.9/lib/python/site-packages (fr
om pandas>=0.18->altair) (2022.7.1)
Requirement already satisfied: MarkupSafe>=2.0 in /Users/vero/Library/Python/3.9/lib/python/site-packages
(from iinia2->altair) (2.1.2)
Requirement already satisfied: six>=1.5 in /Library/Developer/CommandLineTools/Library/Frameworks/Python3.
framework/Versions/3.9/lib/python3.9/site-packages (from python-dateutil>=2.8.1->pandas>=0.18->altair) (1.
15.0)
```

[notice] A new release of pip available: 22.3.1 -> 23.1 [notice] To update, run: /Library/Developer/CommandLineTools/usr/bin/python3 -m pip install --upgrade pip

!pip3 install geopandas In [3]:

hwk5

Defaulting to user installation because normal site-packages is not writeable

Requirement already satisfied: geopandas in /Users/vero/Library/Python/3.9/lib/python/site-packages (0.12.2)

Requirement already satisfied: fiona>=1.8 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from geopandas) (1.9.3)

Requirement already satisfied: pyproj>=2.6.1.post1 in /Users/vero/Library/Python/3.9/lib/python/site-packa ges (from geopandas) (3.5.0)

Requirement already satisfied: pandas>=1.0.0 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from geopandas) (1.5.3)

Requirement already satisfied: shapely>=1.7 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from geopandas) (2.0.1)

Requirement already satisfied: packaging in /Users/vero/Library/Python/3.9/lib/python/site-packages (from geopandas) (23.0)

Requirement already satisfied: cligj>=0.5 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from fiona>=1.8->geopandas) (0.7.2)

Requirement already satisfied: importlib-metadata in /Users/vero/Library/Python/3.9/lib/python/site-packag es (from fiona>=1.8->geopandas) (6.0.0)

Requirement already satisfied: click~=8.0 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from fiona>=1.8->geopandas) (8.1.3)

Requirement already satisfied: munch>=2.3.2 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from fiona>=1.8->qeopandas) (2.5.0)

Requirement already satisfied: attrs>=19.2.0 in /Users/vero/Library/Python/3.9/lib/python/site-packages (f rom fiona>=1.8->geopandas) (22.2.0)

Requirement already satisfied: click-plugins>=1.0 in /Users/vero/Library/Python/3.9/lib/python/site-packag es (from fiona>=1.8->geopandas) (1.1.1)

Requirement already satisfied: certifi in /Users/vero/Library/Python/3.9/lib/python/site-packages (from fi ona>=1.8->geopandas) (2022.12.7)

Requirement already satisfied: numpy>=1.20.3 in /Users/vero/Library/Python/3.9/lib/python/site-packages (f rom pandas>=1.0.0->geopandas) (1.24.1)

Requirement already satisfied: python-dateutil>=2.8.1 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from pandas>=1.0.0->geopandas) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in /Users/vero/Library/Python/3.9/lib/python/site-packages (fr om pandas>=1.0.0->geopandas) (2022.7.1)

Requirement already satisfied: six in /Library/Developer/CommandLineTools/Library/Frameworks/Python3.frame work/Versions/3.9/lib/python3.9/site-packages (from munch>=2.3.2->fiona>=1.8->geopandas) (1.15.0)

Requirement already satisfied: zipp>=0.5 in /Users/vero/Library/Python/3.9/lib/python/site-packages (from importlib-metadata->fiona>=1.8->geopandas) (3.11.0)

[notice] A new release of pip available: 22.3.1 -> 23.1

[notice] To update, run: /Library/Developer/CommandLineTools/usr/bin/python3 -m pip install --upgrade pip

```
In [4]: # importing packages

# data wrangling
import pandas as pd
import numpy as np

# plotting
import altair as alt

# working with geographic data
import json
import geopandas as gpd
from vega_datasets import data
```

Loading in a shape file

This zip file contains a *shape* file. This is one of the types of geometry data supported by geopandas (doc)

```
In [5]: # importing shapefile - We can use a zip file that contains a shape file by reading it from a URL
gdf = gpd.read_file('nps_boundary.zip')
In [6]: gdf.sample(5)
```

Out[6]:

:		UNIT_CODE	GIS_Notes	UNIT_NAME	DATE_EDIT	STATE	REGION	GNIS_ID	UNIT_TYPE	CREATED
	401	HAFE	Lands - http://landsnet.nps.gov/tractsnet/docu	Harpers Ferry National Historical Park	2020-09- 29	WV	NC	1556735	National Historical Park	Lŧ
	371	WHSA	Lands - http://landsnet.nps.gov/tractsnet/docu	White Sands National Park	2020-01- 22	NM	IM	914261	National Park	Lŧ
	203	TUIN	Lands - http://landsnet.nps.gov/tractsnet/docu	Tuskegee Institute National Historic Site	2016-01-14	AL	SE	158362	National Historic Site	Lŧ
	21	вона	Lands - http://landsnet.nps.gov/tractsnet/docu	Boston Harbor Islands National Recreation Area	2006-08- 14	МА	NE	606990	National Recreation Area	Lŧ
	384	TUSK	Lands - http://landsnet.nps.gov/tractsnet/docu	Tule Springs Fossil Beds National Monument	2020-08- 31	NV	PW	2767392	National Monument	Lŧ

Q1 - Overlay the National Parks and Monuments on a county map of the USA

Using the geopandas data frame <code>gdf</code>, filter out just the national parks and monuments and overlay them on a county map of the USA. Use the Vega data set to map the counties, using this page as a reference. Create a tool tip that shows the park name (<code>UNIT_NAME</code>) and region (<code>REGION</code>) of the park, title your chart, and encode the color with the Region of the park using the <code>dark2</code> colorscheme. See this page for a reference on changing the color scheme.

This would be an appropriate visualization if our user wanted to understand the landsize of the parks and how the regions were labeled.

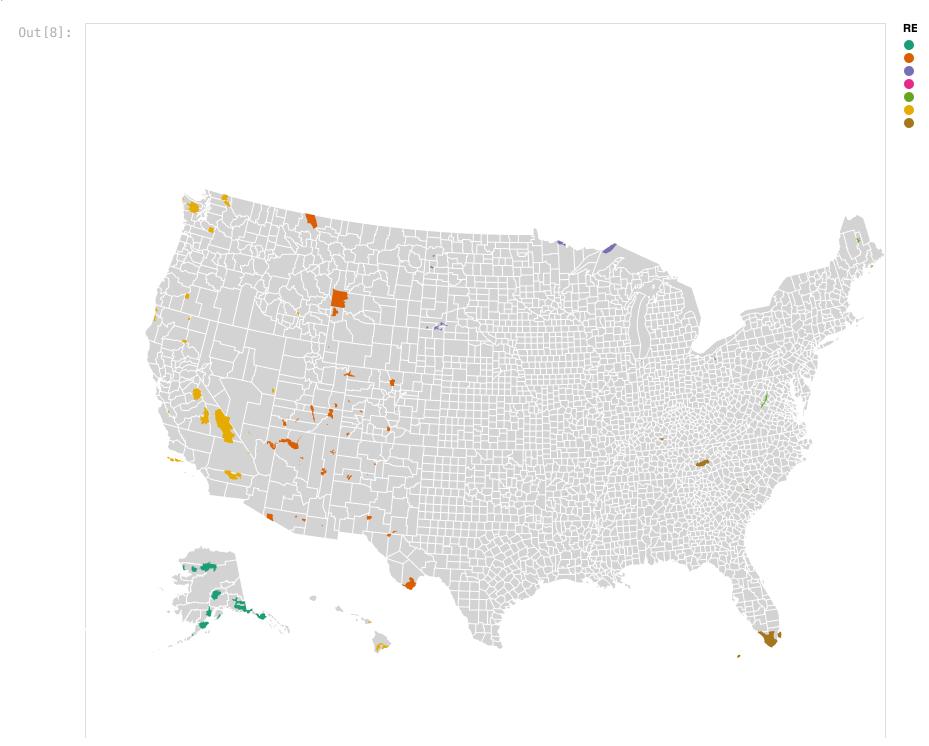
```
In [7]: gdf = gdf[(gdf['UNIT_TYPE'] == 'National Monument') | (gdf['UNIT_TYPE'] == 'National Park')]
```

gdf.head()

Out[7]:		UNIT_CODE	GIS_Notes	UNIT_NAME	DATE_EDIT	STATE	REGION	GNIS_ID	UNIT_TYPE	CREATED_B
	2	LIBI	Lands - http://landsnet.nps.gov/tractsnet/docu	Little Bighorn Battlefield National Monument	2005-02- 23	МТ	IM	806297	National Monument	Lanc
	3	CAVO	Lands - http://landsnet.nps.gov/tractsnet/docu	Capulin Volcano National Monument	2004-07- 22	NM	IM	936867	National Monument	Lanc
	5	GICL	Lands - http://landsnet.nps.gov/tractsnet/docu	Gila Cliff Dwellings National Monument	2004-05- 10	NM	IM	928945	National Monument	Lanc
	6	TUZI	Lands - http://landsnet.nps.gov/tractsnet/docu	Tuzigoot National Monument	2003-11-21	AZ	IM	35601	National Monument	Lanc
	8	CASA	Lands - http://landsnet.nps.gov/tractsnet/docu	Castillo de San Marcos National Monument	2005-02- 28	FL	SE	307911	National Monument	Lanc
In [8]:	fro cou cou	<pre>nties = al ntiesgraph fill='lig stroke='w</pre>	casets import data .t.topo_feature(data.us_10m.url n = alt.Chart(counties).mark_ge phtgray', white' .bersUsa').properties(')					

```
basemap = alt.Chart(gdf).mark_geoshape(
).encode(
    tooltip = ['UNIT_NAME', 'REGION'],
    color=alt.Color('REGION', scale = alt.Scale(scheme='dark2'))
).properties(width=800, height=800)

q1 = countiesgraph + basemap
q1
```



Loading in a csv a species data

```
In [9]: df_species = pd.read_csv('numspecies.csv')
    df_species = df_species.fillna( value = 0 ) # Fill the missing species counts with zeros for visualization
    df_species.sample(5)
```

Out[9]:		UNIT_CODE	All Types	Amphibian	Bird	Fish	Mammal	Reptile	Vascular Plant	Fungi	Insect	Invertebrate	Spider/Scorpion	Nonva
	8	CARE	1566	6.0	248	15.0	73	21.0	1203	0.0	0.0	0.0	0.0	
	42	PEFO	853	8.0	244	0.0	62	21.0	516	1.0	0.0	0.0	0.0	
	11	CONG	2321	43.0	200	65.0	39	50.0	884	279.0	617.0	15.0	21.0	
	9	CAVE	1536	15.0	367	5.0	92	62.0	995	0.0	0.0	0.0	0.0	
	44	REDW	6310	33.0	501	247.0	154	39.0	2257	1363.0	744.0	334.0	7.0	

Q2 - Change data to Long format

Remember long form vs wide form formats? It turns out our species data is wide form. Convert it to long form and save the new dataframe to df_species_long. Name your new columns Species Type and Species Count so they look pretty when we plot them later.

```
In [10]: # your code here
df_species_long = df_species.melt('UNIT_CODE', var_name = 'Species Type', value_name = 'Species Count')
In [11]: # sanity check
df_species_long.sample(5)
```

Out[11]:		UNIT_CODE	Species Type	Species Count		
	453	BLCA	Insect	0.0		
	662	SAGU	Nonvascular Plant	0.0		
	755	HALE	Crab/Lobster/Shrimp	18.0		
	554	VOYA	Invertebrate	7.0		
	540	LAVO	Invertebrate	18.0		

Q3 - Get lat/long and area from park shape

We have the shape file for the national parks, which contains the polygon for the park boundary, but we decide for the visualization we want to encode the location with a circle whose size is constant. We also are interested in comparing the area of the park with the species diversity to discover trends. To accomplish these tasks we need to calculate the centroid of the park to get a single lat/long and calculate the area of the polygon in acres (a convenient unit for the USA).

To do math in geopandas you have to translate between "EPSG:4269" and "EPSG:5070" (USA Albers projection)

Add the Acres column to the gdf geopandas dataframe, and replace the geometry column with the centroid. Use the intro to geopandas documentation to help with the syntax.

Check that everything looks OK by plotting the lat/long of the parks on top of the county map as before. Include a tooltip that shows the region, name of park, and Acres. Use the dark2 color scheme.

I'm helping you out with the coordinate reference systems here; note that there are many many map reference coordinates and we can see the reference for a given geopandas dataframe with the crs property.

In [12]: # look at the coordinate reference of the data — this needs to be translated to do math, but then put back gdf.crs

```
Out[12]: <Geographic 2D CRS: EPSG:4269>
         Name: NAD83
         Axis Info [ellipsoidal]:
         - Lat[north]: Geodetic latitude (degree)
         - Lon[east]: Geodetic longitude (degree)
         Area of Use:

    name: North America - onshore and offshore: Canada - Alberta; British Columbia; Manitoba; New Brunswick;

         Newfoundland and Labrador: Northwest Territories: Nova Scotia: Nunavut: Ontario: Prince Edward Island: Oue
         bec; Saskatchewan; Yukon. Puerto Rico. United States (USA) - Alabama; Alaska; Arizona; Arkansas; Californi
         a; Colorado; Connecticut; Delaware; Florida; Georgia; Hawaii; Idaho; Illinois; Indiana; Iowa; Kansas; Kent
         ucky; Louisiana; Maine; Maryland; Massachusetts; Michigan; Minnesota; Mississippi; Missouri; Montana; Nebr
         aska; Nevada; New Hampshire; New Jersey; New Mexico; New York; North Carolina; North Dakota; Ohio; Oklahom
         a; Oregon; Pennsylvania; Rhode Island; South Carolina; South Dakota; Tennessee; Texas; Utah; Vermont; Virg
         inia; Washington; West Virginia; Wisconsin; Wyoming. US Virgin Islands. British Virgin Islands.
         - bounds: (167.65, 14.92, -40.73, 86.45)
         Datum: North American Datum 1983
         - Ellipsoid: GRS 1980
         - Prime Meridian: Greenwich
In [13]: | qdf = qdf.to crs( 'EPSG:5070')
In [14]: gdf.crs
```

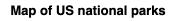
```
Out[14]: <Projected CRS: EPSG:5070>
         Name: NAD83 / Conus Albers
         Axis Info [cartesian]:
         - X[east]: Easting (metre)
         - Y[north]: Northing (metre)
         Area of Use:

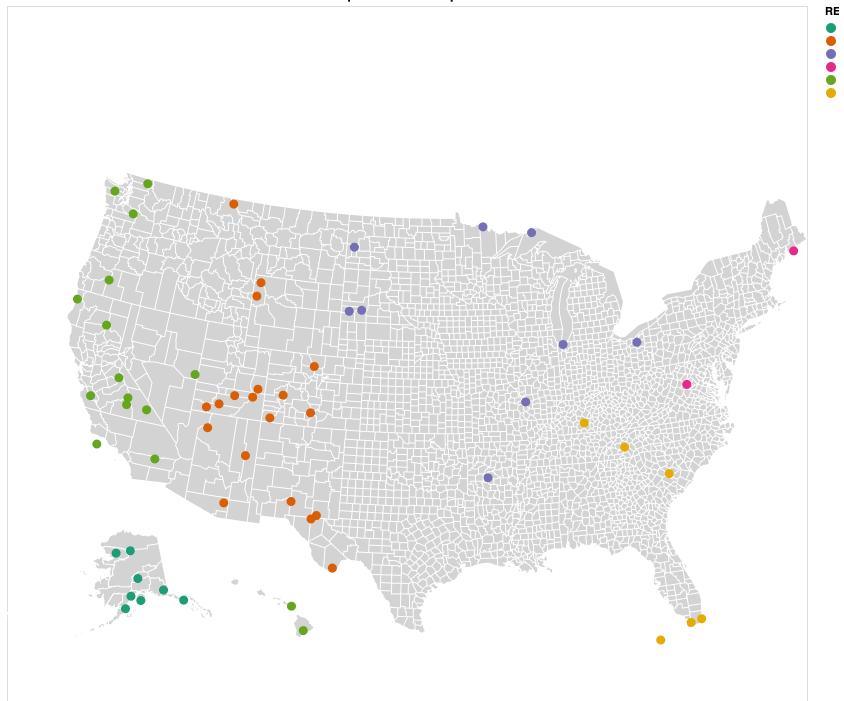
    name: United States (USA) - CONUS onshore - Alabama; Arizona; Arkansas; California; Colorado; Connecticu

         t; Delaware; Florida; Georgia; Idaho; Illinois; Indiana; Iowa; Kansas; Kentucky; Louisiana; Maine; Marylan
         d; Massachusetts; Michigan; Minnesota; Mississippi; Missouri; Montana; Nebraska; Nevada; New Hampshire; Ne
         w Jersey; New Mexico; New York; North Carolina; North Dakota; Ohio; Oklahoma; Oregon; Pennsylvania; Rhode
         Island; South Carolina; South Dakota; Tennessee; Texas; Utah; Vermont; Virginia; Washington; West Virgini
         a; Wisconsin; Wyoming.
         - bounds: (-124.79, 24.41, -66.91, 49.38)
         Coordinate Operation:
         - name: Conus Albers
         - method: Albers Equal Area
         Datum: North American Datum 1983
         - Ellipsoid: GRS 1980
         - Prime Meridian: Greenwich
In [15]: # Find the area in acres
         gdf['Acres'] = gdf.area / 4047
In [16]: # sanity check
         # The value from Rocky Mountain National Park webpage is 265,807 acres and our estimate is off probably due
         # projection we used was not locally optimal, or maybe the shape file wasn't exact. If we were a GIS class
         # this would probably bother us, but for now we'll call this close enough (your number should be about 0.59
         qdf[qdf['UNIT CODE'] == 'ROMO']['Acres']
Out[16]: 415
                267032.175621
         Name: Acres, dtype: float64
In [17]: # your code here to convert the polygons data to point data
         gdf['geometry'] = gdf.centroid
In [18]: qdf = qdf.to crs( 'EPSG:4269') # This takes us back to lat/long space
         gdf.sample(5)
```

Out[18]:	UNI ⁻	T_CODE	GIS_Notes	UNIT_NAME	DATE_EDIT	STATE	REGION	GNIS_ID	UNIT_TYPE	CREATED
	160	CAGR	Lands - http://landsnet.nps.gov/tractsnet/docu	Casa Grande Ruins National Monument	2020-04- 29	AZ	IM	44811	National Monument	Le
	171	ORCA	Lands - http://landsnet.nps.gov/tractsnet/docu	Oregon Caves National Monument and Preserve	2015-02- 26	OR	PW	1147260	National Monument	Le
	51	SEQU	Lands - http://landsnet.nps.gov/tractsnet/docu	Sequoia National Park	2008-11-26	CA	PW	266000	National Park	Le
	27	ELMO	Lands - http://landsnet.nps.gov/tractsnet/docu	El Morro National Monument	2006-12-11	NM	IM	906062	National Monument	La
	101	JEFF	Lands - http://landsnet.nps.gov/tractsnet/docu	Gateway Arch National Park	2010-10-28	МО	MW	765817	National Park	La
In [19]:).encode tool cold).proper q3 = cou q3 = q3.	e(tip = pr=alt.(rties(wintiesg) proper	Chart(gdf[gdf['UNIT_TYPE'] == ' ['UNIT_NAME', 'REGION', 'Acres' Color('REGION', scale = alt.Sca idth=800, height=800) raph + basemap ties(ap of US national parks'	1,		_geosha	ape(

Out[19]:





Q4 - Merge Geo Data and Species Data

We've been using the UNIT_NAME for the park name in the plots, and we want to use that same name, and the acerage when plotting the species data. Add these two columns to our df_species_long dataframe using a inner join, renaming the UNIT_NAME to Park Name, and REGION to Region for pretty visualization. The key for the merge is UNIT_CODE. Save this new dataframe as df_species_final

```
In [20]: # your code here
df_species_final = gdf.merge(df_species_long, how = 'inner', on = 'UNIT_CODE')
df_species_final = df_species_final.rename(columns = {'UNIT_NAME': 'Park Name', 'REGION': 'Region'})
df_species_final
```

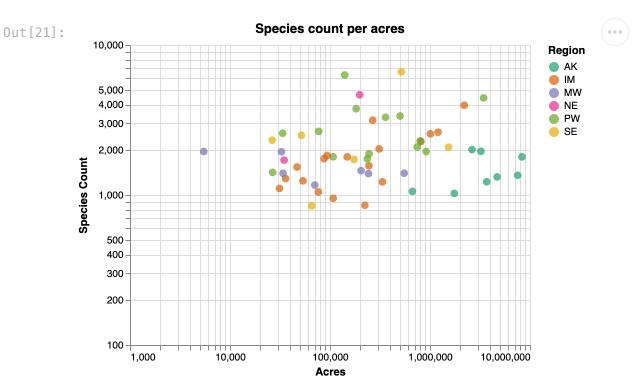
Out[20]:	UNI	r_code	GIS_Notes	Park Name	DATE_EDIT	STATE	Region	GNIS_ID	UNIT_TYPE	CREATED_BY
_	0	DRTO	Lands - http://landsnet.nps.gov/tractsnet/docu	Dry Tortugas National Park	2007-11-07	FL	SE	307632	National Park	Lands
	1	DRTO	Lands - http://landsnet.nps.gov/tractsnet/docu	Dry Tortugas National Park	2007-11-07	FL	SE	307632	National Park	Lands
	2	DRTO	Lands - http://landsnet.nps.gov/tractsnet/docu	Dry Tortugas National Park	2007-11-07	FL	SE	307632	National Park	Lands
	3	DRTO	Lands - http://landsnet.nps.gov/tractsnet/docu	Dry Tortugas National Park	2007-11-07	FL	SE	307632	National Park	Lands
	4	DRTO	Lands - http://landsnet.nps.gov/tractsnet/docu	Dry Tortugas National Park	2007-11-07	FL	SE	307632	National Park	Lands
	•••									
	820	SAGU	Lands - http://landsnet.nps.gov/tractsnet/docu	Saguaro National Park	2021-03- 29	AZ	IM	10683	National Park	Lands
	821	SAGU	Lands - http://landsnet.nps.gov/tractsnet/docu	Saguaro National Park	2021-03- 29	AZ	IM	10683	National Park	Lands
	822	SAGU	Lands - http://landsnet.nps.gov/tractsnet/docu	Saguaro National Park	2021-03- 29	AZ	IM	10683	National Park	Lands
	823	SAGU	Lands - http://landsnet.nps.gov/tractsnet/docu	Saguaro National Park	2021-03- 29	AZ	IM	10683	National Park	Lands
	824	SAGU	Lands - http://landsnet.nps.gov/tractsnet/docu	Saguaro National Park	2021-03- 29	AZ	IM	10683	National Park	Lands

825 rows x 19 columns

Q5 - Compare the Acres to Species Count

On a log/log scale compare the total species count (only the rows where "Species Type" is "All Types") with the park size. Encode the region using the same color scale as above. Add a tool tip to see the park name, Species Count, and Acres.

Add a selection_multi that will highlight the selected data by turning unselected data light gray.



Q6 - Create a rug plot of Species Diversity

Look at how the different Species categores (all Species types that are NOT 'All Types' are distributed across all the parks.

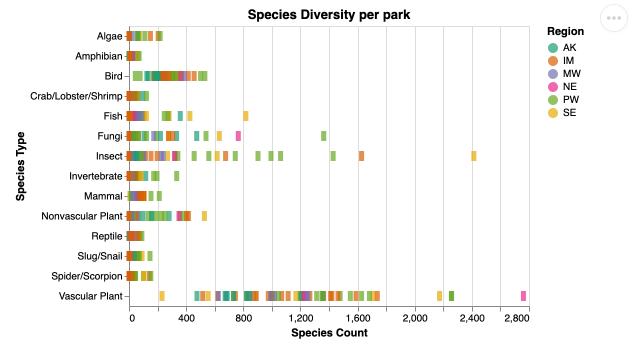
Use the same color encoding for the Region as the prior plots.

Add a selection_multi that will highlight the selected data for a park by turning unselected data light gray. Also, make the tick a little bigger when it is selected to help find the park in the other categories.

```
color = alt.condition(selection, alt.Color('Region:N', scale = alt.Scale(scheme = 'dark2')), alt.value
tooltip = ['Park Name'],
    size = alt.condition(selection, alt.value(10), alt.value(2))
).properties(
    title = 'Species Diversity per park'
)

q6.add_selection(
    selection
)
```

Out[22]:



Q7 - Create a sorted bar chart of Species Count with a Region Drop down selector

Create a bar chart with horizontal bars that shows the total species count per park with the color channel encoding the Region.

Add a drop down selector so that you can filter the bar chart by just one region. See this documentation for an example of adding a drop down menu and using it to filter your selections.

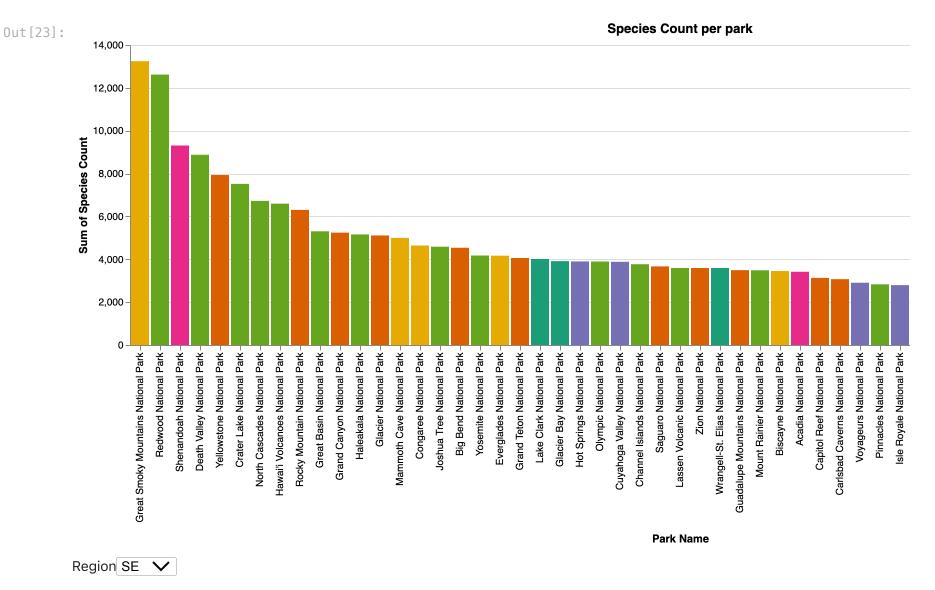
In addition to the drop down, include a selection_multi that will highlight the selected data by turning unselected data light gray.

```
In [23]: # your code here

regions = list(df_species_final['Region'].unique())
dropdown = alt.binding_select(options = regions, name = "Region")
selection = alt.selection_single(fields=['Region'], bind=dropdown)

q7 = alt.Chart(df_species_final).mark_bar().encode(
    x = alt.X('Park Name', sort = '-y'),
    y = alt.Y('sum(Species Count)'),
    color = alt.condition(
    selection,
    alt.Color('Region:0', scale = alt.Scale(scheme = 'dark2')),
    alt.value('lightgray'))
).properties(
    title = 'Species Count per park')

q7.add_selection(
    selection
)
```



Q8 - Bring all the charts together for a Dashboard

Bring all 4 charts and the region filter together. Any click on one chart should highlight that data in the other charts. The colors for the regions should not change depending on the region selection and be consistent across all graphs. When plotting your geopandas data frame, filter out only the rows that have species data before making the map. I used the

4/16/23, 1:00 PM hwk5

documentation on customizing titles to change the font size and name my dashboard (all of the titles in the component graphs I changed to be subtitles)

Change your color scheme to tableau10.

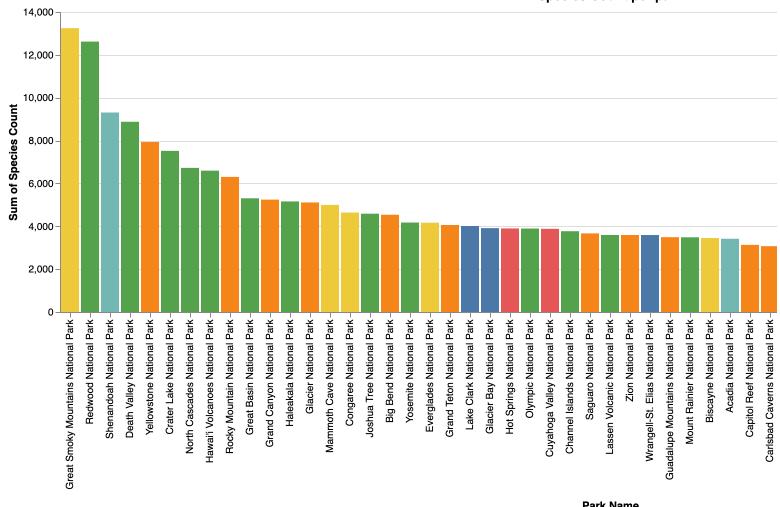
Hint: There will be only one selection_multi that links all the selection highlighting in the charts together.

```
In [24]: # regions = list(df species final['Region'].unique())
         # labels = [region + ' ' for region in regions]
         # dropdown = alt.binding_select(options = regions + [None], labels = labels + ['All'], name = "Region")
         # selector = alt.selection single(fields=['Region'], bind=dropdown)
         selector = alt.selection multi(fields = ['Region'])
         color = alt.condition(selector, alt.Color('Region:0', scale = alt.Scale(scheme = 'tableau10')),
                               alt.value('lightgray'))
         countiesgraph = alt.Chart(counties).mark_geoshape(
             fill='lightgray',
             stroke='white'
         ).project('albersUsa').properties(
             width=800,
             height=800
         basemap = alt.Chart(df_species_final).mark_geoshape(
         ) encode(
             tooltip = ['Park Name', 'Region', 'Acres'],
             color=color
         ).properties(width=800, height=800)
         q3 = countiesgraph + basemap
         q3 = q3.add selection(
         selector).properties(
             title = 'Map of US national parks'
         q5 = alt.Chart(df_species_all).mark_circle(size = 60).encode(
             x = alt.X('Acres:0', scale = alt.Scale(type = 'log')),
             y = alt.Y('Species Count:Q', scale = alt.Scale(type = 'log')),
             color = color,
             tooltip=['Park Name', 'Species Count', 'Acres']
```

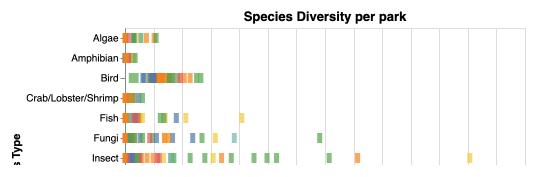
```
).add_selection(
selector).properties(
    title = 'Species count per acres'
q6 = alt.Chart(species_diff).mark_tick(size = 15, thickness = 5).encode(
   x = 'Species Count',
   y = 'Species Type',
    color = color,
    tooltip = ['Park Name'],
    size = alt.condition(selector, alt.value(10), alt.value(2))
).add selection(
selector
).properties(
    title = 'Species Diversity per park'
q7 = alt.Chart(df_species_final).mark_bar().encode(
    x = alt.X('Park Name', sort = '-y'),
    y = alt.Y('sum(Species Count)'),
    color = color
).add selection(
selector).properties(
    title = 'Species Count per park'
q8 = q7 & q6 & q5 & q3
q8 = q8.properties(
    title = 'National Parks and Biodiversity Demographics Dashboard'
8p
```

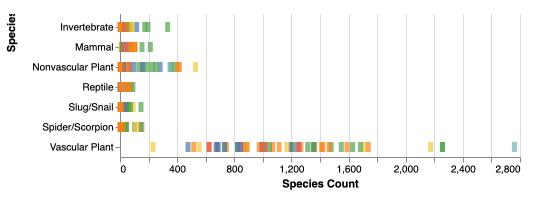
National Parks and Biodiversity Demographics Dashboard

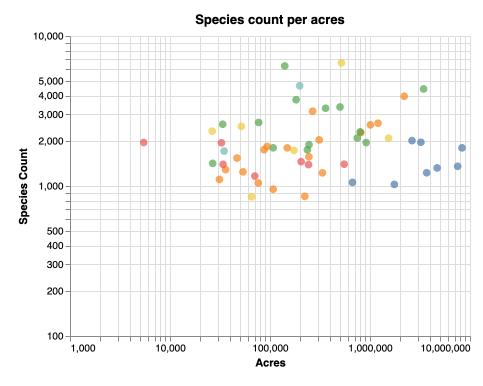




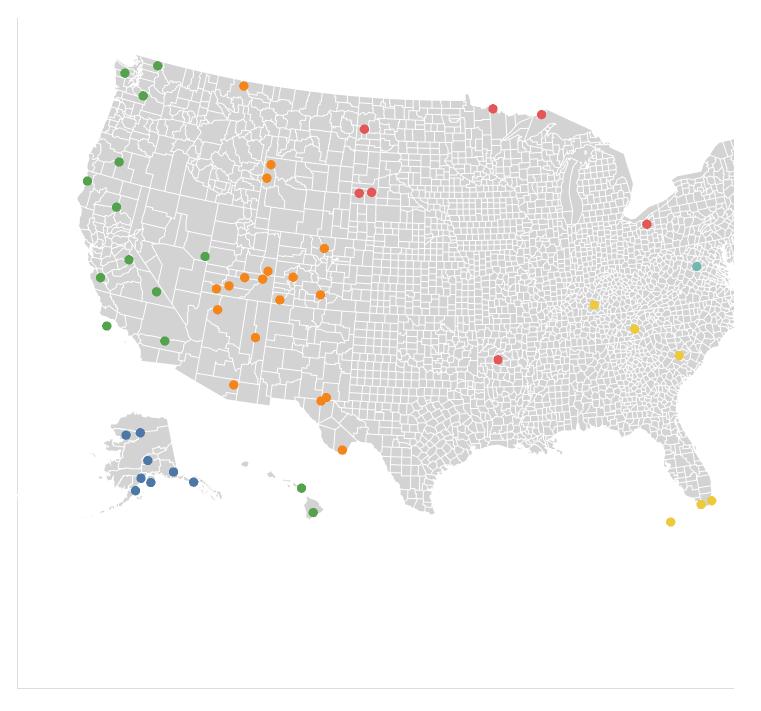








Map of US national parks



4/16/23, 1:00 PM hwk5

In []:	
In []:	