Lab 5 Oregon Fires

Lauren Ponisio

Conservation/ecology Topics

- Explore how Oregon fires are changing due to fire suppression and climate change.
- Describe fundamental concepts in fire ecology, including fire severity.

Statistical Topics

- Describe the fundamental attributes of a raster dataset.

Computational Topics

- Explore raster attributes and metadata using R.
- Import rasters into R using the `terra` package.
- Plot raster files in R using the `ggplot2` package.
- Reproject raster and vector data
- Layer raster and vector data together

Lab part 1: reading in fire raster data and plotting

We will be working with the soil burn severity data from the 2020 Holiday Farm Fire (up the McKenzie E of Eugene), the 2020 Beachie Fire (near Portland) and the 2018 Terwilliger fire (up the McKenzie E of Eugene, near Cougar hotsprings).

We will use data downloaded from the USGS: https://burnseverity.cr.usgs.gov/products/baer

Specifically, BARC Fire Severity layers are created by first calculating spectral indices from pre- and post-fire satellite imagery that are sensitive to changes caused by fire. The two images are then subtracted showing the difference between them which is then binned into 4 burn severity classes (high, moderate, low, very low/unburned). Field crews ground-truth the severity classes.

The metadata files provide additional details on how the continuous data was binned into discrete catagories.

a. Read in each fire severity rasters, name them [fire name]_rast. The .tif files are the rasters.

HINT: The files are nested within folders so be aware of your file paths.

```
setwd("~/ds-environ-aw/5-OR-fires")
rebel_sbs2_rast <-
  rast("soil-burn-severity/2017_rebel_sbs/rebel_sbs2.tif")
```

```
SoilSeverity_rast <-
  rast("soil-burn-severity/2018_terwilliger_sbs/SoilSeverity.tif")

BeachieCreek_SBS_final_rast <-
  rast("soil-burn-severity/2020_beachiecreek_sbs/BeachieCreek_SBS_final.tif")


HolidayFarm_SBS_final_rast <-
  rast("soil-burn-severity/2020_holidayfarm_sbs/HolidayFarm_SBS_final.tif")
```

b. Summarize the values of the rasters. Take note of the labels associated with the data values because you will need it for plotting.

```
summary(rebel_sbs2_rast)
```

```
## Warning: [summary] used a sample


##      Layer_1
## 15      :96149
## 2       : 3257
## 3       :  504
## 1       :  209
## 4       :  135
## 0       :    0
## (Other):    0
```

```
summary(SoilSeverity_rast)
```

```
## Warning: [summary] used a sample


##      SoilBurnSe
## Unburned: 8801
## Low      :25507
## Moderate: 4337
## High     :  586
## NA's     :61113
```

```
summary(BeachieCreek_SBS_final_rast)
```

```
## Warning: [summary] used a sample

##      Layer_1
## 3       :23275
## 2       :14608
## 4       : 4623
## 1       : 1969
## 0       :    0
## (Other):    0
## NA's    :55625
```

2

```r
summary(HolidayFarm_SBS_final_rast)
```

```
## Warning: [summary] used a sample
```

```
##      Layer_1
## 3      :33163
## 2      :12950
## 4      : 4933
## 1      : 2305
## 0      :    0
## (Other):    0
## NA's   :46837
```

    c. Plot each raster.. Set the scale to be `scale_fill_brewer(palette = "Spectral", direction=-1)`

HINT: Remember we have to turn them into "data.frames" for ggplot to recognize them as plot-able.

HINT HINT: Remember to check the labels of the data values to be able to set the fill.

```r
holiday_df <- as.data.frame(HolidayFarm_SBS_final_rast, xy = TRUE)
str(holiday_df)
```

```
## 'data.frame':    1750770 obs. of  3 variables:
##  $ x      : num  541078 541098 541118 541138 541158 ...
##  $ y      : num  4903324 4903324 4903324 4903324 4903324 ...
##  $ Layer_1: Factor w/ 127 levels "0","1","2","3",..: 2 2 2 2 2 2 2 2 2 2 ...
```

```r
ggplot() +
    geom_raster(data = holiday_df, aes(x = x, y = y, fill = Layer_1)) +
  scale_fill_brewer(palette = "Spectral", direction =-1)
```

```
## Warning: Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using `geom_tile()` instead.
```

```r
beachie_df <- as.data.frame(BeachieCreek_SBS_final_rast, xy = TRUE)
str(beachie_df)
```

```
## 'data.frame':    1948877 obs. of  3 variables:
##  $ x      : num  395819 395839 395859 395879 395899 ...
##  $ y      : num  1221232 1221232 1221232 1221232 1221232 ...
##  $ Layer_1: Factor w/ 127 levels "0","1","2","3",..: 3 3 3 3 3 3 3 3 3 3 ...
```

```r
ggplot() +
    geom_raster(data = beachie_df, aes(x = x, y = y, fill = Layer_1)) +
  scale_fill_brewer(palette = "Spectral", direction =-1)
```

```
## Warning: Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using `geom_tile()` instead.
```
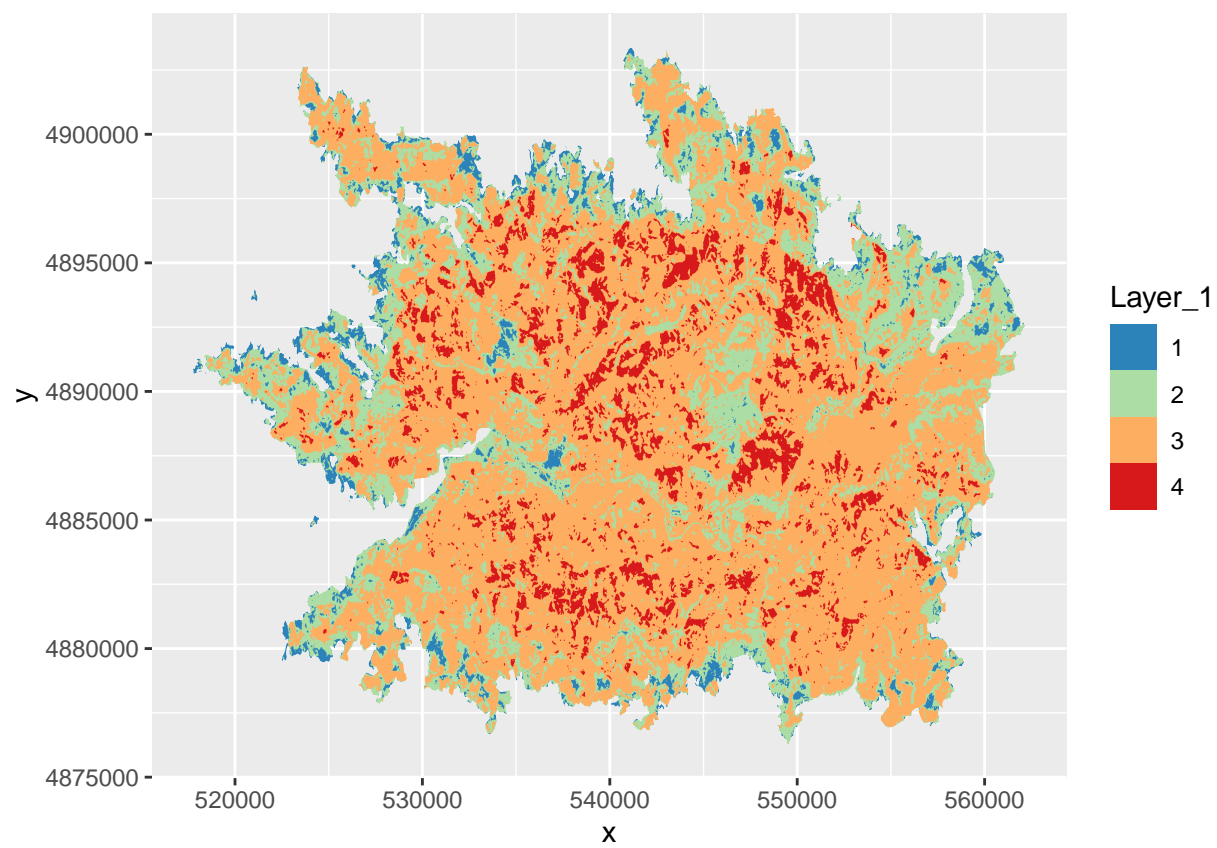
Figure 1: Holiday plot with ggplot2 using the Spectral color scale
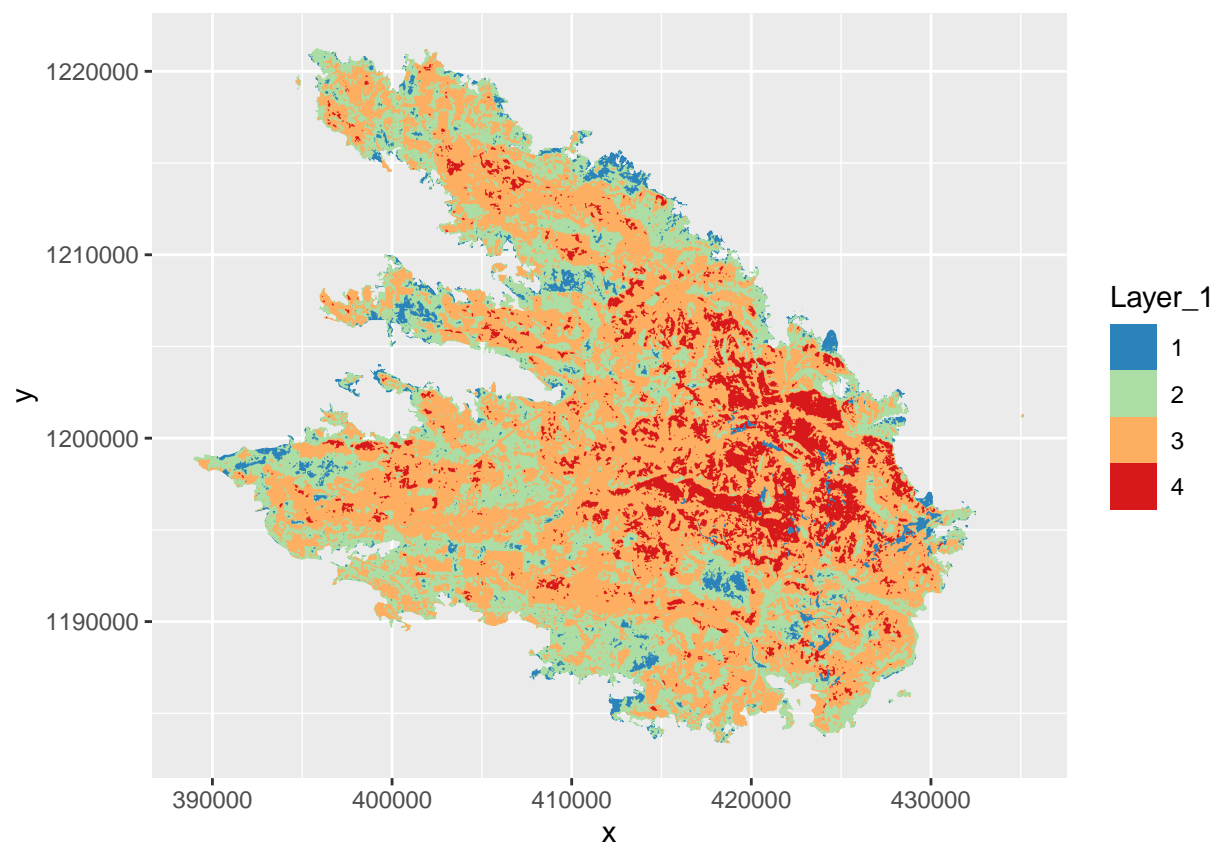
Figure 2: Beachie plot with ggplot2 using the Spectral color scale

```
terwilliger_df <- as.data.frame(SoilSeverity_rast, xy = TRUE)
str(terwilliger_df)
```

```
## 'data.frame':    51544 obs. of  3 variables:
##  $ x        : num  419860 419890 419920 419950 420340 ...
##  $ y        : num  1119923 1119923 1119923 1119923 1119923 ...
##  $ SoilBurnSe: Factor w/ 4 levels "Unburned","Low",..: 1 2 1 2 2 2 1 2 1 1 ...
```

```
ggplot() +
    geom_raster(data = terwilliger_df, aes(x = x, y = y, fill = SoilBurnSe)) +
  scale_fill_brewer(palette = "Spectral", direction =-1)
```
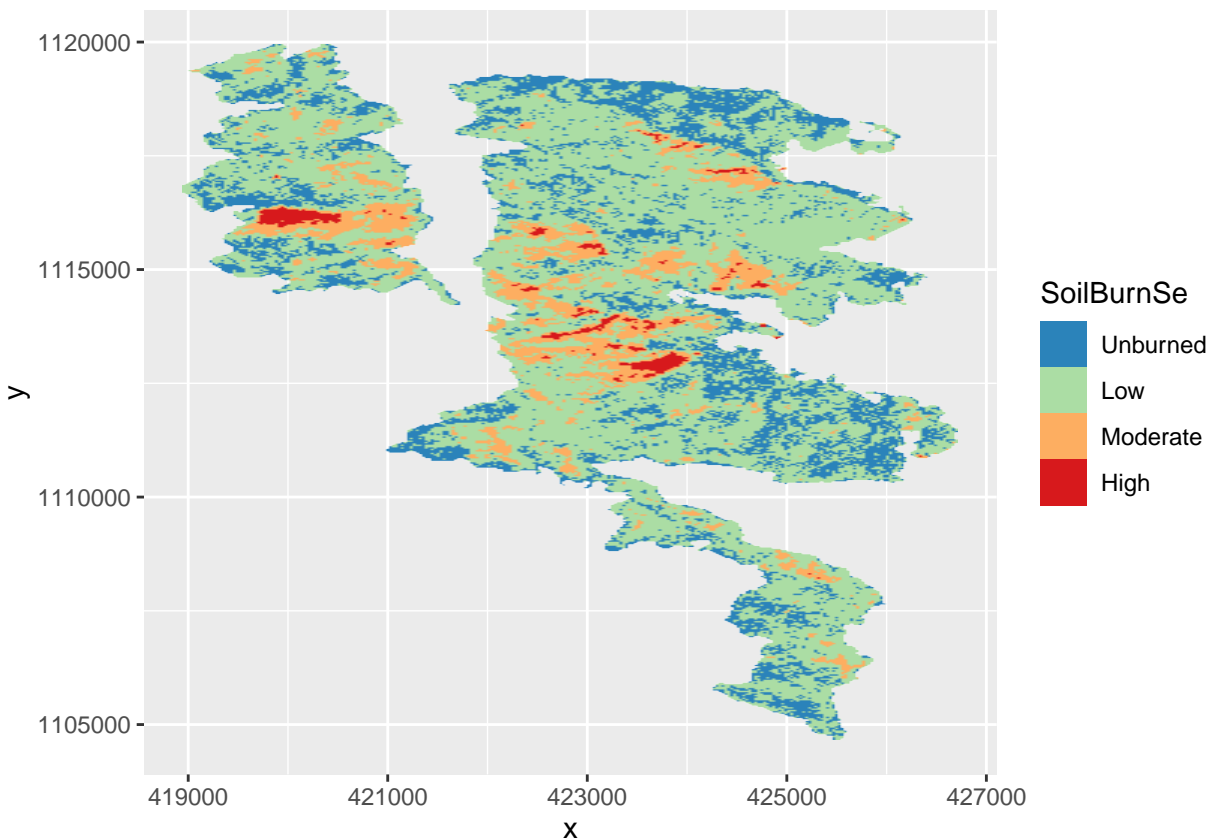

Figure 3: Terwilliger plot with ggplot2 using the Spectral color scale

d. Compare these visualizations what is something you notice? -ANSWER: Looking at all of the visualizations, the first thing I noticed was that the terwilliger fire had the least amount of 3 and 4 level severity overall.

Lab part 2: Exploring the attributes of our spatial data.

a. What are the crs of the rasters? What are the units? Are they all the same?

```r
crs(HolidayFarm_SBS_final_rast, proj = TRUE)
```

```
## [1] "+proj=utm +zone=10 +datum=NAD83 +units=m +no_defs"
```

```r
crs(BeachieCreek_SBS_final_rast, proj = TRUE)
```

```
## [1] "+proj=aea +lat_0=34 +lon_0=-120 +lat_1=43 +lat_2=48 +x_0=600000 +y_0=0 +datum=NAD83 +units=m +n
```

```r
crs(SoilSeverity_rast, proj = TRUE)
```

```
## [1] "+proj=aea +lat_0=34 +lon_0=-120 +lat_1=43 +lat_2=48 +x_0=600000 +y_0=0 +datum=NAD83 +units=m +n
```

- ANSWER crs: Holiday: utm Beachie: aea Terwilliger: aea
- ANSWER units: Holiday: m Beachie: m Terwilliger: m
- ANSWER the same? : Beachie and Terwilliger are the same, Holiday is different crs.

b. What about the resolution of each raster?

```r
res(HolidayFarm_SBS_final_rast)
```

```
## [1] 20 20
```

```r
res(BeachieCreek_SBS_final_rast)
```

```
## [1] 20 20
```

```r
res(SoilSeverity_rast)
```

```
## [1] 30 30
```

- ANSWER resolution: Holiday: 20 20 Beachie: 20 20 Terwilliger: 30 30
- ANSWER the same? : No they are all different.

c. Calculate the min and max values of each raster. Are they all the same?

```r
minmax(HolidayFarm_SBS_final_rast)
```

```
##      Layer_1
## min        1
## max      127
```

```r
minmax(BeachieCreek_SBS_final_rast)
```

```
##      Layer_1
## min        1
## max      127
```

```
minmax(SoilSeverity_rast)
```

```
##      SoilBurnSe
## min         1
## max         4
```

- ANSWER minmax: Holiday: min: 1 max: 127 Beachie: min: 1 max: 127 Terwilliger: min: 1 max:4
- ANSWER the same? : The minimimum values are all the same but the max value for terwilliger is much lower than Holiday and Beachie.

Given we expect there to be 4 values for each bin of severity (high, moderate, low, very low/unburned), let's try to work out why there are values other than 1-4. After checking the metadata .txt and inspecting the metadata in the raster itself, I could not find an explicit mention of the meaning on the non 1-4 data (maybe you can?). Not great practices USGS! But it is likely missing data. Let's convert the Holiday data greater than 4 to NA, just like we would a regular matrix of data.

```
HolidayFarm_SBS_final_rast[HolidayFarm_SBS_final_rast > 4] <- NA
summary(values(HolidayFarm_SBS_final_rast))
```

```
##      Layer_1
##  Min.   :1.0
##  1st Qu.:2.0
##  Median :3.0
##  Mean   :2.8
##  3rd Qu.:3.0
##  Max.   :4.0
##  NA's   :1536190
```

That's better :)

d. Do the same conversion for Beachie.

```
BeachieCreek_SBS_final_rast[BeachieCreek_SBS_final_rast > 4] <- NA
summary(values(BeachieCreek_SBS_final_rast))
```

```
##      Layer_1
##  Min.   :1.0
##  1st Qu.:2.0
##  Median :3.0
##  Mean   :2.7
##  3rd Qu.:3.0
##  Max.   :4.0
##  NA's   :2437627
```

Lab part 3: Reprojection

From our exploration above, the rasters are not in the same projection, so we will need to re-project them if we are going to be able to plot them together.

We can use the `project()` function to reproject a raster into a new CRS. The syntax is `project(RasterObject, crs)`

a. First we will reproject our `beachie_rast` raster data to match the `holidat_rast` CRS. If the resolution is different, change it to match Holiday's resolution.

Don't change the name from beachie_rast.

```r
beachie_rast <- project(BeachieCreek_SBS_final_rast, crs(HolidayFarm_SBS_final_rast))

#Fixing complexity of the names I made earlier
holiday_rast <- HolidayFarm_SBS_final_rast
terwilliger_rast <- SoilSeverity_rast


# This should return TRUE
crs(beachie_rast, proj = TRUE) == crs(holiday_rast, proj = TRUE)
```

```
## [1] TRUE
```

b. Now convert the Terwilliger crs to the holiday crs. If the resolution is different, change it to match Holiday's resolution.

```r
terwilliger_rast <- project(terwilliger_rast, crs(holiday_rast), res = res(holiday_rast))

# This should return TRUE TRUE
crs(terwilliger_rast, proj = TRUE) == crs(holiday_rast, proj = TRUE)
```

```
## [1] TRUE
```

```r
res(terwilliger_rast)[2] == res(holiday_rast)[2]
```

```
## [1] TRUE
```

c. Now you can plot all of the fires on the same map! HINT: Remember to re-make the dataframes.

```r
terwilliger_df <- as.data.frame(terwilliger_rast, xy = TRUE)
holiday_df <- as.data.frame(holiday_rast, xy = TRUE)
beachie_df <- as.data.frame(beachie_rast, xy = TRUE)

ggplot() +
    geom_raster(data = terwilliger_df,
                aes(x = x, y = y,
                 fill = SoilBurnSe)) +
    geom_raster(data = holiday_df,
                aes(x = x, y = y,
                 fill = Layer_1)) +
      geom_raster(data = beachie_df,
                aes(x = x, y = y,
                 fill = Layer_1)) +
    scale_fill_brewer(palette = "Spectral", direction =-1)
```

```
## Warning: Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using `geom_tile()` instead.
## Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using `geom_tile()` instead.
```
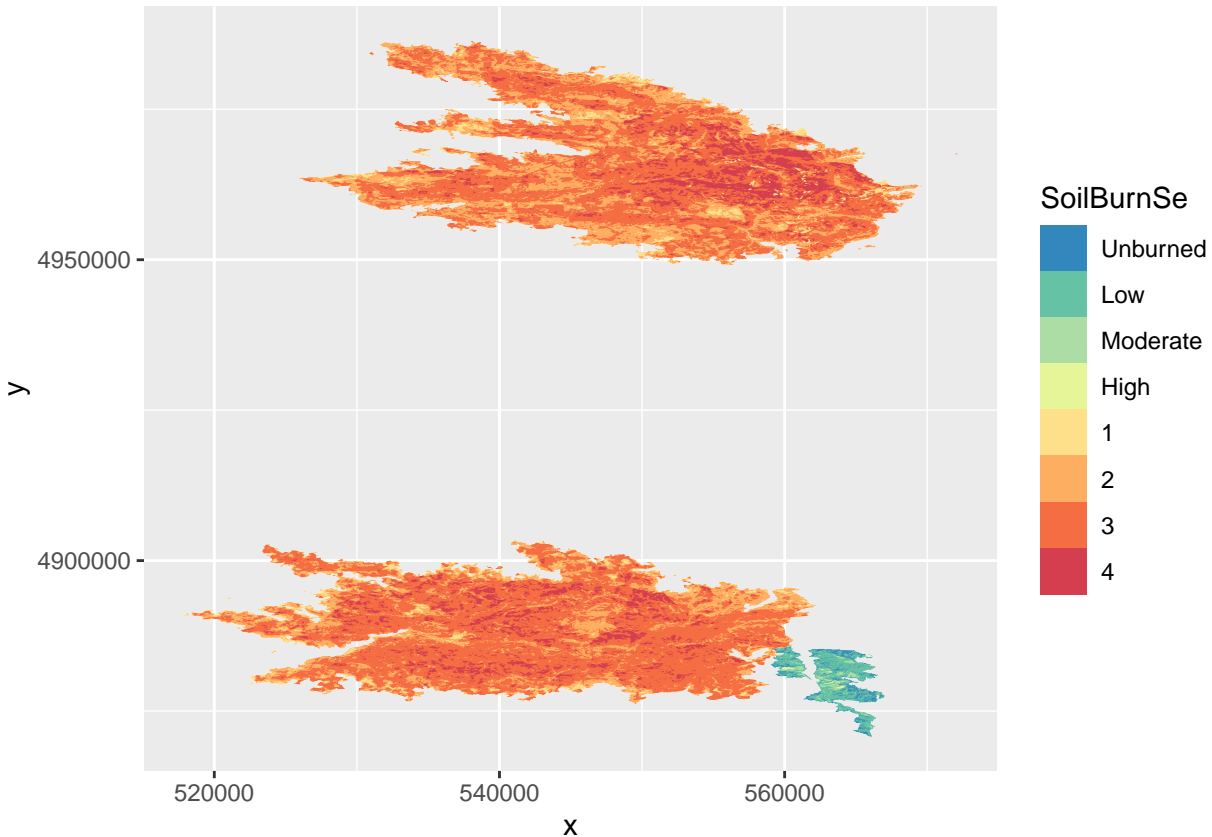
Well that's annoying. It appears as though in 2018 the makers of these data decided to give 1,2,3,4 categorical names which are being interpreted as two different scales. If we look at the terwilliger_rast values we can see that in min max.

```
terwilliger_rast$SoilBurnSe
```

```
## class       : SpatRaster
## dimensions  : 776, 417, 1  (nrow, ncol, nlyr)
## resolution  : 20, 20  (x, y)
## extent      : 558901, 567241, 4870585, 4886105  (xmin, xmax, ymin, ymax)
## coord. ref. : NAD83 / UTM zone 10N (EPSG:26910)
## source(s)   : memory
## categories  : SoilBurnSe, BAER_Acres
## name        : SoilBurnSe
## min value   :    Unburned
## max value   :        High
```

d. Let's deal with the the easy way and modify the dataframe. Convert High to 4, Moderate to 3, Low to 2, and Unburned to 1 using your data subsetting skills.

Somethings you will need to be careful of: - If you check the class of terwilliger_rast_df$SoilBurnSe it is a factor, which is a special class of data that are ordered categories with specific levels. R will not let you convert add a level. So first, convert the data to characters (using as.character()). - Now the data are characters, so you will not be able to add in numerics. So code the 1,2,3 as characters i.e., "1", "2"... - We will eventually want the data to be factors again so it will match up with the other rasters. So lastly, convert the data to a factor (using as.factor()).

```r
terwilliger_df$SoilBurnSe <- as.character(terwilliger_df$SoilBurnSe)

terwilliger_df$SoilBurnSe[terwilliger_df$SoilBurnSe == "High"] <- "4"
terwilliger_df$SoilBurnSe[terwilliger_df$SoilBurnSe == "Moderate"] <- "3"
terwilliger_df$SoilBurnSe[terwilliger_df$SoilBurnSe == "Low"] <- "2"
terwilliger_df$SoilBurnSe[terwilliger_df$SoilBurnSe == "Unburned"] <- "1"

terwilliger_df$SoilBurnSe <- as.factor(terwilliger_df$SoilBurnSe)
```

e. Try plotting again.

```r
ggplot() +
    geom_raster(data = terwilliger_df,
                aes(x = x, y = y,
                 fill = SoilBurnSe)) +
    geom_raster(data = holiday_df,
                aes(x = x, y = y,
                 fill = Layer_1)) +
      geom_raster(data = beachie_df,
                aes(x = x, y = y,
                 fill = Layer_1)) +
    scale_fill_brewer(palette = "Spectral", direction =-1)
```
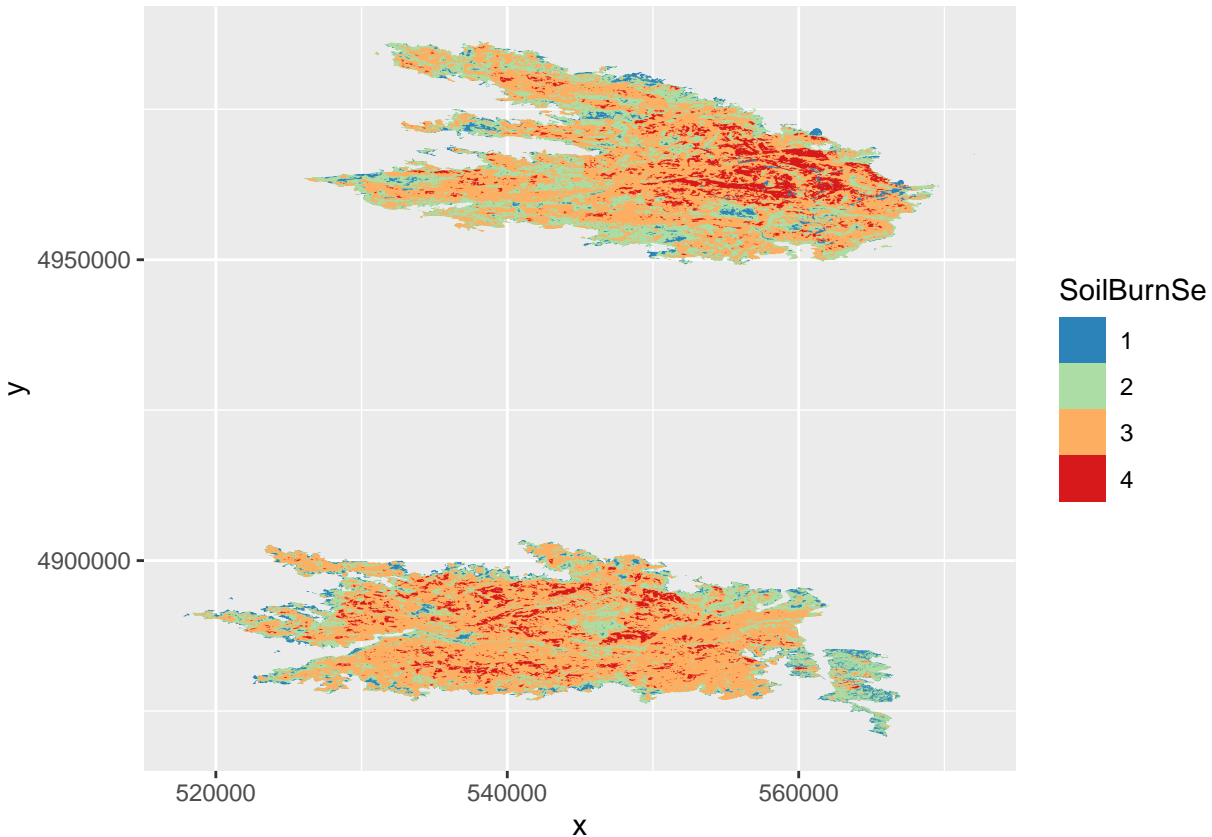
```
## Warning: Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using 'geom_tile()' instead.
## Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using 'geom_tile()' instead.
```

The scale bar make sense! It would be nice to have a baselayer map to see where is Oregon these fires are.

Lab part 4: Adding in vector data

I found a nice ecoregion map on the OR spatial data website. https://spatialdata.oregonexplorer.info/geoportal/details;id=3c7862c4ae664993ad1531907b1e413e

    a. Load the data into R, it is in the OR-ecoregions folder.

```
ecoregions <- st_read(
  "OR-ecoregions/Ecoregions_OregonConservationStrategy.shp")
```

```
## Reading layer 'Ecoregions_OregonConservationStrategy' from data source
##   '/Users/avawessel/ds-environ-aw/5-OR-fires/OR-ecoregions/Ecoregions_OregonConservationStrategy.shp
##   using driver 'ESRI Shapefile'
## Simple feature collection with 9 features and 6 fields
## Geometry type: POLYGON
## Dimension:     XY
## Bounding box:  xmin: 183871.7 ymin: 88600.88 xmax: 2345213 ymax: 1675043
## Projected CRS: NAD83 / Oregon GIC Lambert (ft)
```

    b. Check the projection and re-project if needed. We did not cover this in the lecture demo, but for vector data, use st_transform()

```r
crs(ecoregions, proj = TRUE)
```

```
## [1] "+proj=lcc +lat_0=41.75 +lon_0=-120.5 +lat_1=43 +lat_2=45.5 +x_0=399999.9999984 +y_0=0 +datum=NAP
```

```r
ecoregions <- st_transform(ecoregions, crs(holiday_rast))
```

c. Plot all of the data together (the rasters and vector data). You can layer on geom_sf into ggplot with the other rasters just like you would add another raster.

```r
ggplot() +
    geom_raster(data = terwilliger_df,
                aes(x = x, y = y,
                 fill = SoilBurnSe)) +
    geom_raster(data = holiday_df,
                aes(x = x, y = y,
                  fill = Layer_1)) +
      geom_raster(data = beachie_df,
                aes(x = x, y = y,
                 fill = Layer_1)) +
    geom_sf(data = ecoregions,
            aes(fill = NA)) +
    scale_fill_brewer(palette = "Spectral", direction =-1)
```

```
## Warning: Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using 'geom_tile()' instead.
## Raster pixels are placed at uneven horizontal intervals and will be shifted
## i Consider using 'geom_tile()' instead.
```
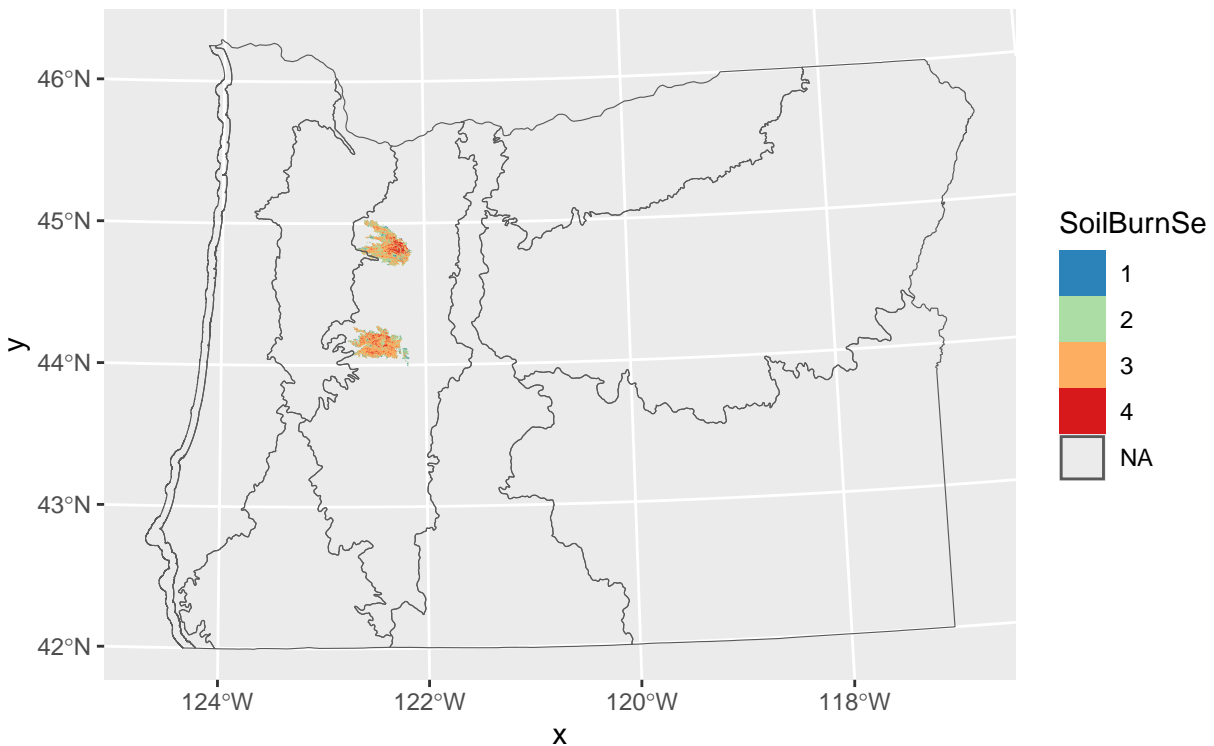
We could get fancy and zoom into the correct region using extent, which we will cover next week. For now, this looks pretty good.

Lab part 5: Exploring patterns of fire severity

a. Create a barplot with the count of each fire severity category.

- Use scale_fill_brewer(palette = "Spectral", direction=-1) to get the bars to match the maps.
- Plot the proportion on the y. To do this, in geom_bar, include y = (..count..)/sum(..count..). EX: aes(x= Layer_1, y = (..count..)/sum(..count..)

HINT: Rather annoyingly, you will need to convert the layer values to factors again to get fill to recognize them. EX: fill=as.factor(Layer_1)

```
library(patchwork)
```

```
##
## Attaching package: 'patchwork'
```

```
## The following object is masked from 'package:terra':
##
##     area
```

```
holiday_plot <- ggplot(data = holiday_df, aes(x = Layer_1, y = (..count..)/sum(..count..), fill = as.fa
  geom_bar() +
  scale_fill_brewer(palette = "Spectral", direction = -1) +
  labs(x = "Fire Severity Category", y = "Proportion", fill = "Severity Level")

beachie_plot <- ggplot(data = beachie_df, aes(x = Layer_1, y = (..count..)/sum(..count..), fill = as.fa
  geom_bar() +
  scale_fill_brewer(palette = "Spectral", direction = -1) +
  labs(x = "Fire Severity Category", y = "Proportion", fill = "Severity Level")

terwilliger_plot <- ggplot(data = terwilliger_df, aes(x = SoilBurnSe, y = (..count..)/sum(..count..), f
  geom_bar() +
  scale_fill_brewer(palette = "Spectral", direction = -1) +
  labs(x = "Fire Severity Category", y = "Proportion", fill = "Severity Level")

combined <- holiday_plot + beachie_plot + terwilliger_plot + plot_layout(ncol = 1)
combined
```
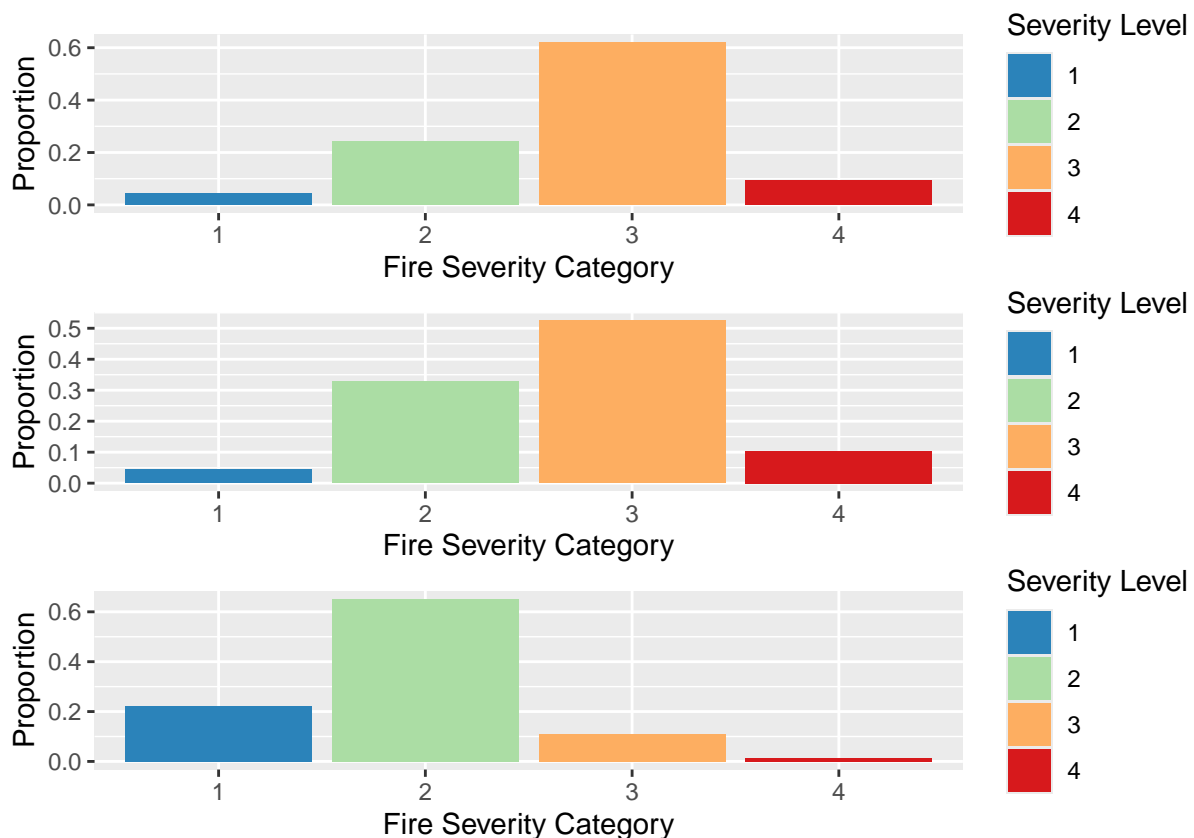
```
## Warning: The dot-dot notation ('..count..') was deprecated in ggplot2 3.4.0.
## i Please use 'after_stat(count)' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```


b. What do you notice about the frequency of different severity classes when you compare these barplots. How does this relate to the Haldofsky reading? ANSWER: Looking at different barplots for each fire, The

15

frequency of level 3 severity makes up over half of the fire for two of the three fires. The second most common category is level two severity, which is pretty high in two of the three fires, one of the fires having this make up more than half. Relating this back to the Haldofsky reading, the researchers discussed heavily about the severity of fires and how this could increase with our changing climate. The researchers stressed that there is a lot of considerations, such as resistant tree logging and droughts, that larger high-severity fires have the potential to increase. This along with increased fire freqency and reburns can result in higher severity levels shown for future fires.

Also, if the legend label bothers you (as it does for me) Check out this tutorial: https://www.datanovia. com/en/blog/ggplot-legend-title-position-and-labels/