



Working with Geospatial Data in R

The raster package

Data frames aren't a great way to store spatial data

```
> head(preds)
  lon lat predicted_price
1 -123.3168 44.52539      258936.2
2 -123.3168 44.52740      257258.4
3 -123.3168 44.52940      255543.1
4 -123.3168 44.53141      253791.0
5 -123.3168 44.53342      252002.4
6 -123.3168 44.53542      250178.7
```

- No CRS information
- Inefficient storage
- Inefficient display

A better structure for raster data

- data matrix + information on grid + CRS

		258936.2	256579.2	254147.2	251593.8	...
25		258936.2	256579.2	254147.2	251593.8	...
25	25	258936.2	256579.2	254147.2	251593.8	...
25	25	257258.4	255082.5	252848.8	250499.2	
	25	255543.1	253557.9	251537.5	249410.6	
		253791.0	252004.4	250211.4	248326.8	
		...				

The `raster` package

- `sp` provides some raster data classes:
 - `SpatialGrid`, `SpatialPixels`,
`SpatialGridDataFrame`, `SpatialPixelsDataFrame`
- But `raster` is better:
 - easier import of rasters
 - large rasters aren't read into memory
 - provides functions for raster type operations
- Also uses S4 and when appropriate provides same functions

raster provides print methods for sp objects

```
> library(sp)
> countries_spdf
```

```
An object of class "SpatialPolygonsDataFrame"
Slot "data":
```

	name	iso_a3	population	gdp
region				
1	Afghanistan	AFG	28400000	22270.00
2	Angola	AGO	12799293	110300.00
3	Albania	ALB	3639453	21810.00

...

VERY long output!

```
Slot "proj4string":
```

```
CRS arguments:
```

```
+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0
```

raster provides print methods for sp objects

```
> library(raster)
> countries_spdf

class          : SpatialPolygonsDataFrame
features       : 177
extent         : -180, 180, -90, 83.64513 (xmin, xmax, ymin, ymax)
coord. ref.    : +proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 ...
variables      : 6
names          :          name, iso_a3, population,          gdp,
min values     : Afghanistan,      -99,          140,          16.00,
max values     :      Zimbabwe,      ZWE, 1338612970, 15094000.00, ...
```

Compact and useful output



Working with Geospatial Data in R

Let's practice!

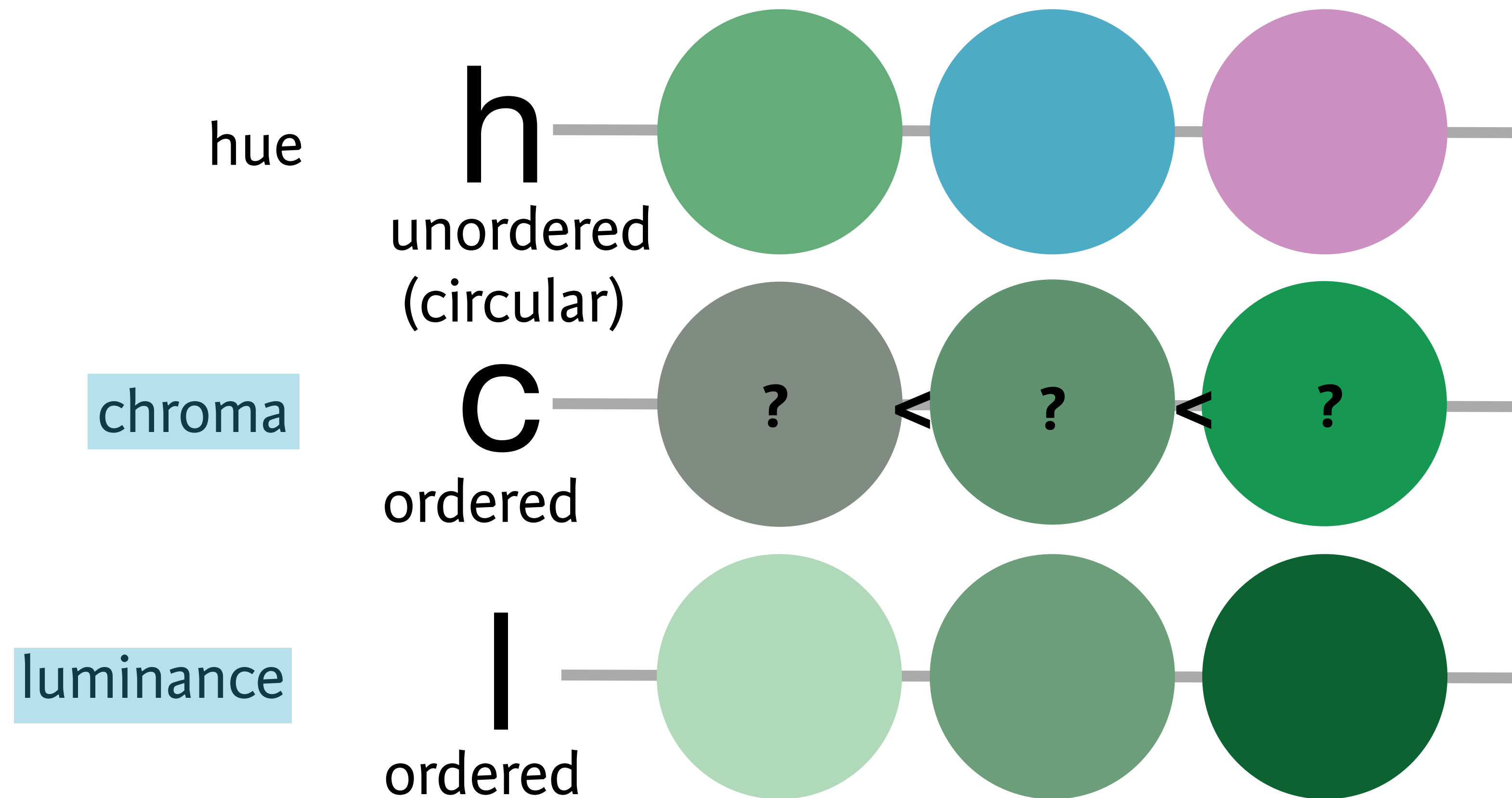


Working with Geospatial Data in R

Color

A perceptual color space: HCL

- **Trichromatic** - we perceive color as three-dimensional



Types of scale

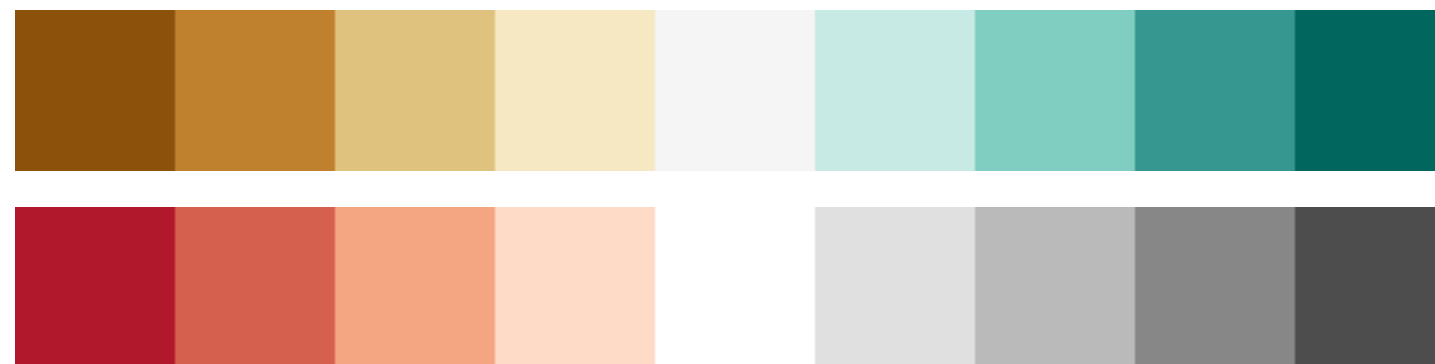
- Sequential - ordered

steps in chroma and/or luminance
hue maybe redundant coding



- Diverging - ordered but in two directions

steps in chroma and/or luminance
with hue distinguishing direction



- Qualitative - unordered

steps in hue with
equal chroma and luminance



Generating color scales in R

```
> library(RColorBrewer)
> display.brewer.all()

> brewer.pal(n = 9, "Blues")
[1] "#F7FBFF" "#DEEBF7" "#C6DBEF" "#9ECAE1"
[5] "#6BAED6" "#4292C6" "#2171B5" "#08519C"
[9] "#08306B"

> library(viridisLite)
> viridis(n = 9)
[1] "#440154FF" "#472D7BFF" "#3B528BFF" "#2C728EFF"
[5] "#21908CFF" "#27AD81FF" "#5DC863FF" "#AADC32FF"
[9] "#FDE725FF"
```

transparency



Working with Geospatial Data in R

Let's practice!

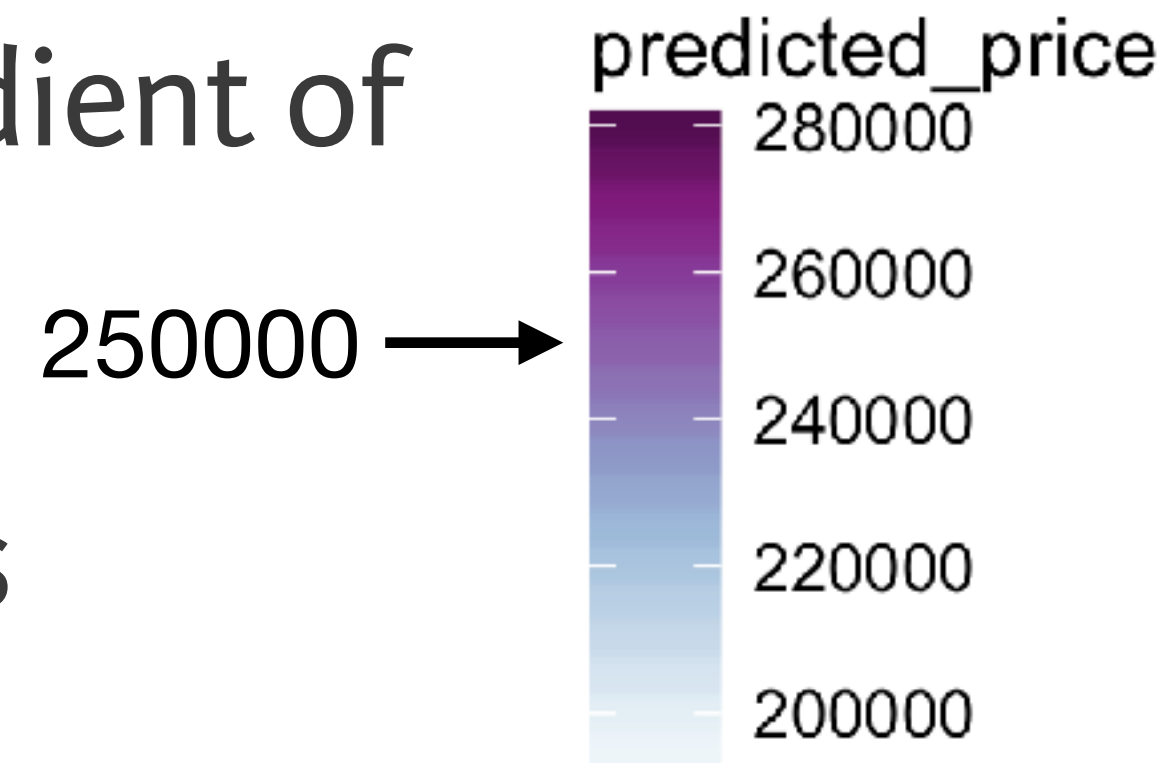


Working with Geospatial Data in R

Color scales 2

Mapping of numbers to color

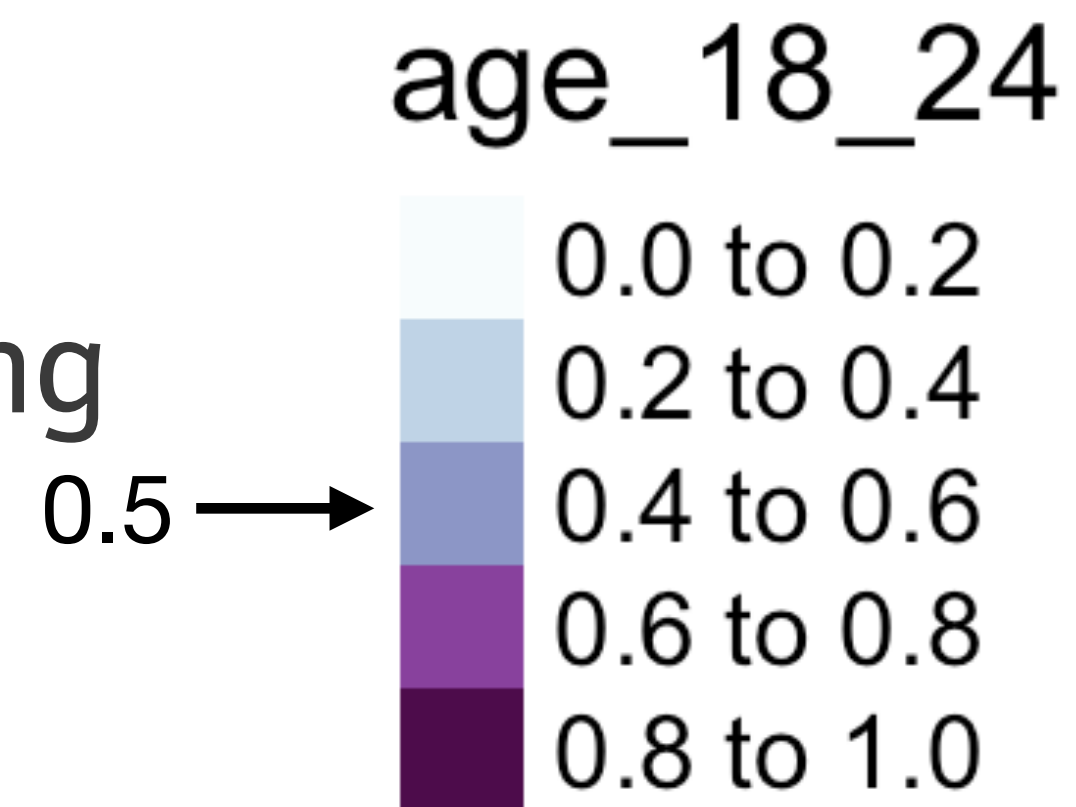
- `ggplot2`: map to a continuous gradient of color



- `tmap`: map to a discrete set of colors

- Continuous map: control mapping by transforming the scale, e.g log

- Discrete map: control mapping by binning the variable



Discrete vs. continuous mapping

- Continuous:
 - Perceptually uniform: perceiving equivalent color difference to numerical difference
- Discrete:
 - Complete control over scale
 - Easier lookup

Cutting a variable into bins

```
> lib  
> cla
```

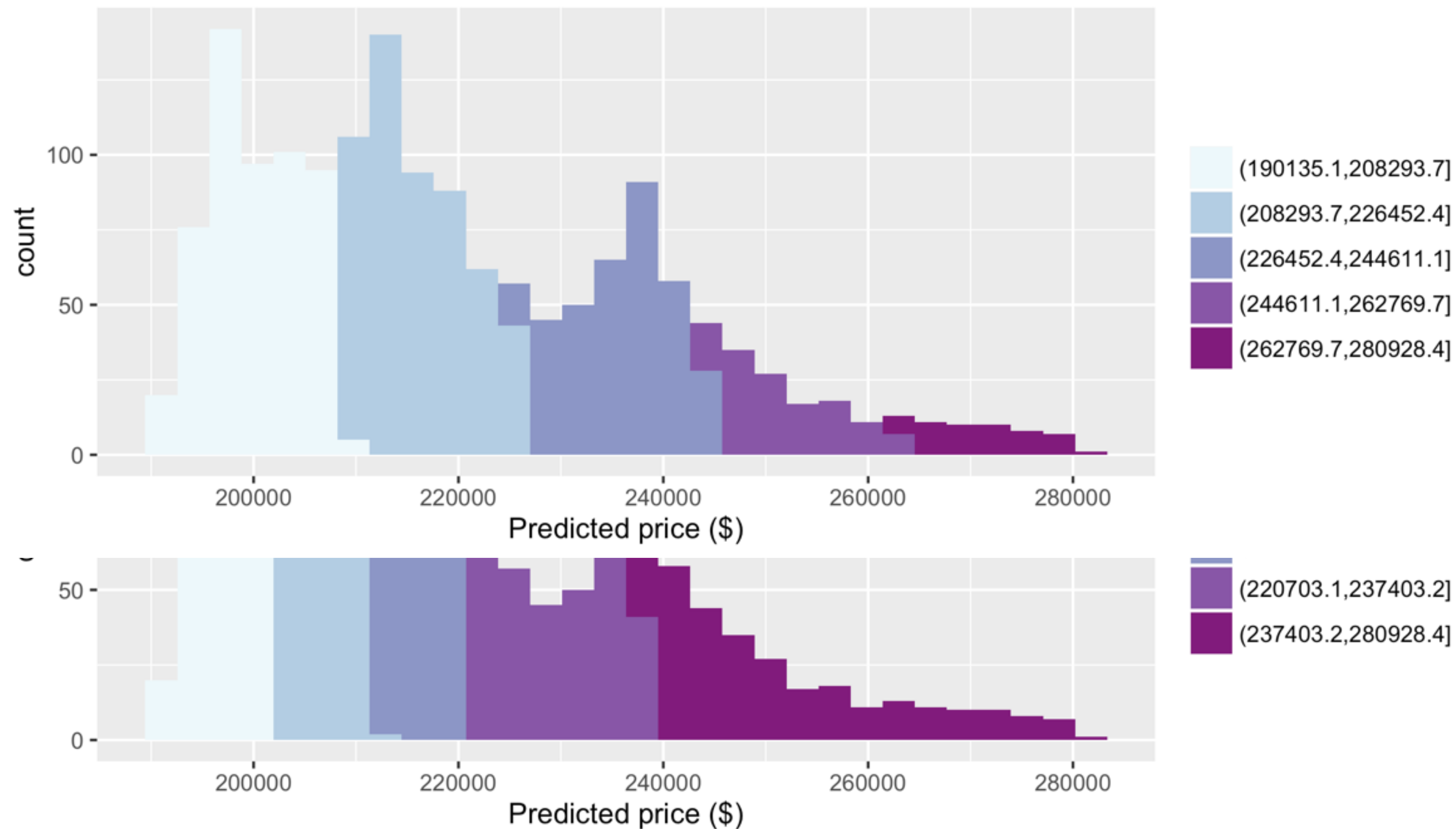
```
style  
[1901
```

```
[2446
```

```
> cla
```

```
style  
[1901
```

```
[2207
```



Cutting a variable into bins

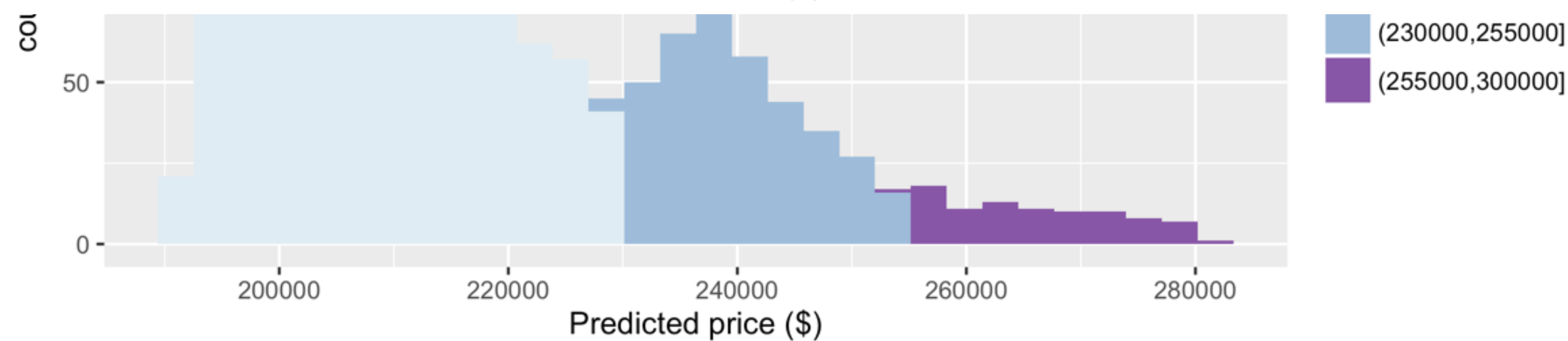
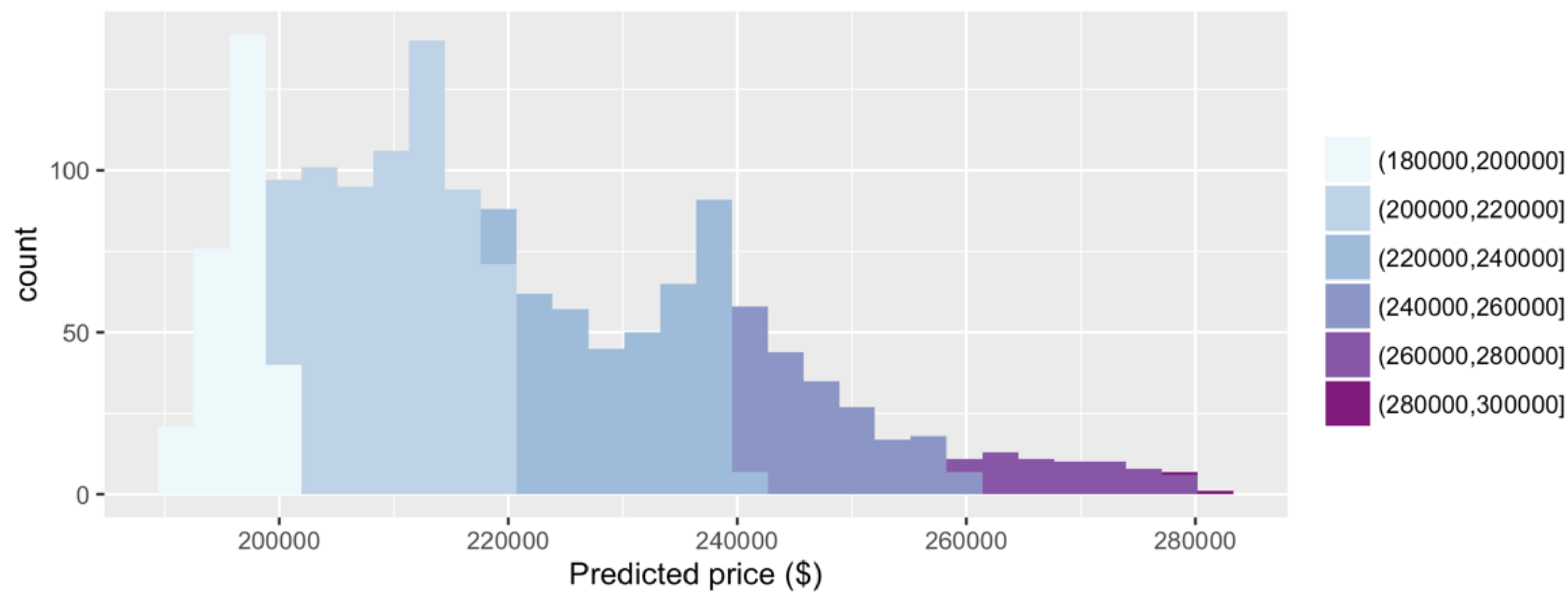
```
> class
```

```
style:  
[18000
```

```
[260000
```

```
> class  
fix
```

```
style:  
[1e+05
```





Working with Geospatial Data in R

Let's practice!