# Operations on Raster Data I

HES 505 Fall 2024: Session 12

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#### **Objectives**

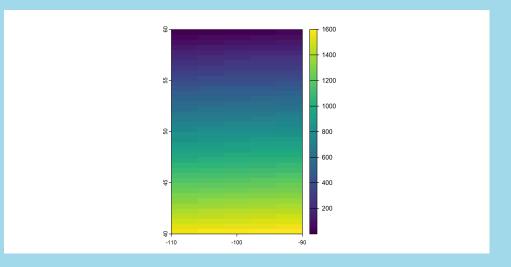
- By the end of today, you should be able to:
  - Align rasters for spatial processing
  - Adjust the resolution of raster data
  - Combine (or reduce) rasters to match the extent of your analysis

# Aligning rasters for spatial processing

#### Projecting raster data

- Transformation from lat/long to planar
   CRS involves some loss of precision
- New cell values estimated using overlap with original cells
- Interpolation for continuous data,
   nearest neighbor for categorical data
- Equal-area projections are preferred; especially for large areas

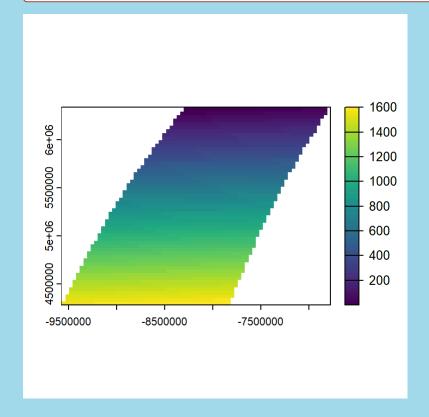
```
1 library(sf)
2 library(terra)
3 library(spDataLarge)
4 r <- rast(xmin=-110, xmax=-90, ymin=40, ym
5 values(r) <- 1:ncell(r)
6 plot(r)</pre>
```



#### Projecting raster data

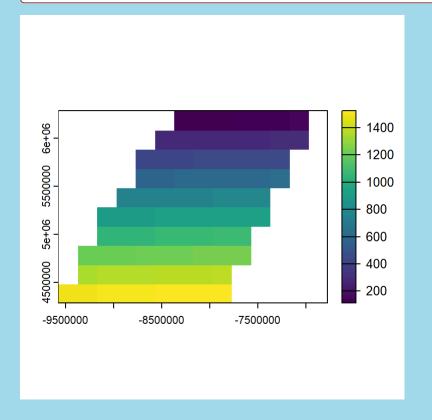
• simple method; alignment • providing a template to not guaranteed

```
1 newcrs <- "+proj=robin +da
 pr1 <- terra::project(r, r</pre>
3 plot(pr1)
```



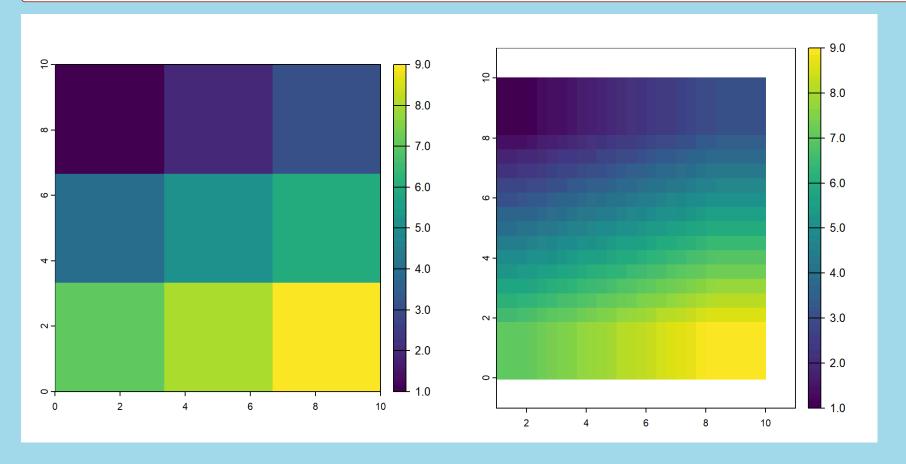
## ensure alignment

```
1 \times < - rast(pr1)
2 # Set the cell size
3 \text{ res}(x) < -200000
4 pr3 <- terra::project(r,</pre>
5 plot(pr3)
```



#### Aligning Data: resample

```
1 r <- rast(nrow=3, ncol=3, xmin=0, xmax=10, ymin=0, ymax=10)
2 values(r) <- 1:ncell(r)
3 s <- rast(nrow=25, ncol=30, xmin=1, xmax=11, ymin=-1, ymax=11)
4 x <- resample(r, s, method="bilinear")</pre>
```



# Adjusting resolution

#### Downscaling and Upscaling

- Aligning data for later analysis
- Remembering scale
- Thinking about support

#### Changing resolutions

- aggregate, disaggregate, resample allow changes in cell size
- aggregate requires a function (e.g., mean() or min()) to determine what to do with the grouped values
- resample allows changes in cell size and shifting of cell centers (slower)

### Changing resolutions: aggregate

1 r <- rast() 2 r

class : SpatRaster

dimensions : 180, 360, 1 (nrow,

ncol, nlyr)

resolution : 1, 1 (x, y)

extent : -180, 180, -90, 90

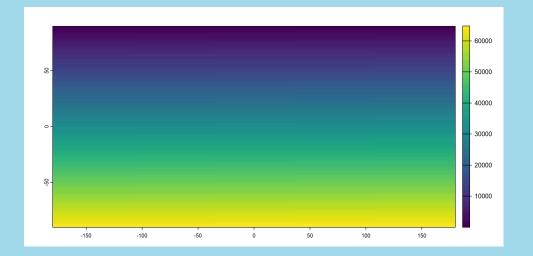
(xmin, xmax, ymin, ymax)

coord. ref. : lon/lat WGS 84 (CRS84)

(OGC:CRS84)

1 values(r) <- 1:ncell(r)</pre>

2 plot(r)



1 ra <- aggregate(r, 20)
2 ra</pre>

class : SpatRaster

dimensions : 9, 18, 1 (nrow, ncol,

nlyr)

resolution : 20, 20 (x, y)

extent : -180, 180, -90, 90

(xmin, xmax, ymin, ymax)

coord. ref. : lon/lat WGS 84 (CRS84)

(OGC:CRS84)

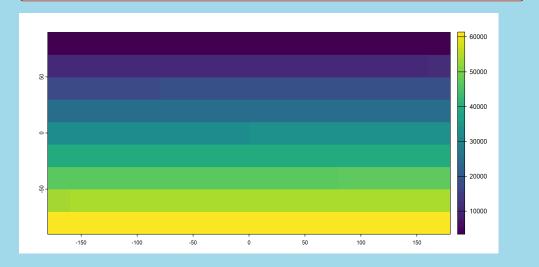
source(s) : memory

name : lyr.1

min value : 3430.5

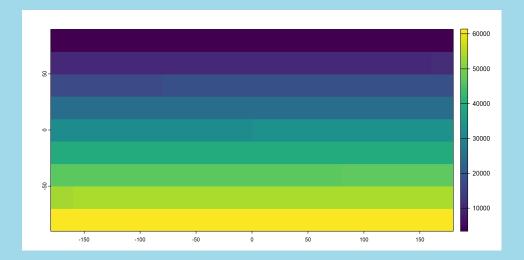
max value : 61370.5

#### 1 plot(ra)



### Changing resolutions: disagg

1 ra <- aggregate(r, 20)
2 plot(ra)</pre>



```
1 rd <- disagg(r, 20)</pre>
```

1 rd

class : SpatRaster

dimensions : 3600, 7200, 1 (nrow,

ncol, nlyr)

resolution : 0.05, 0.05 (x, y) extent : -180, 180, -90, 90

(xmin, xmax, ymin, ymax)

coord. ref.: lon/lat WGS 84 (CRS84)

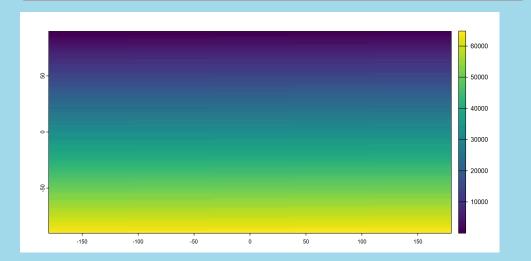
(OGC:CRS84)

source :

spat\_27a03b4b79c4\_10144.tif

name : lyr.1
min value : 1
max value : 64800

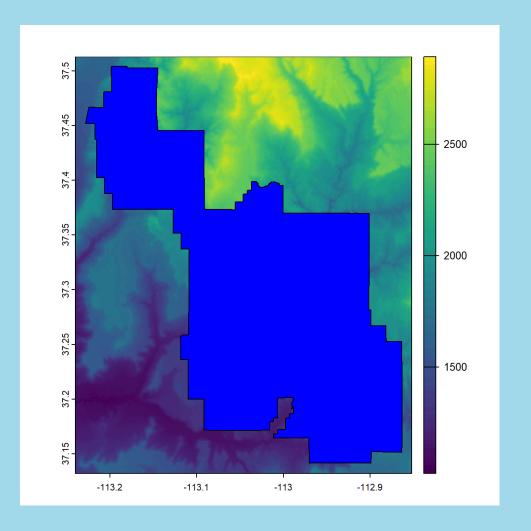
1 plot(rd)



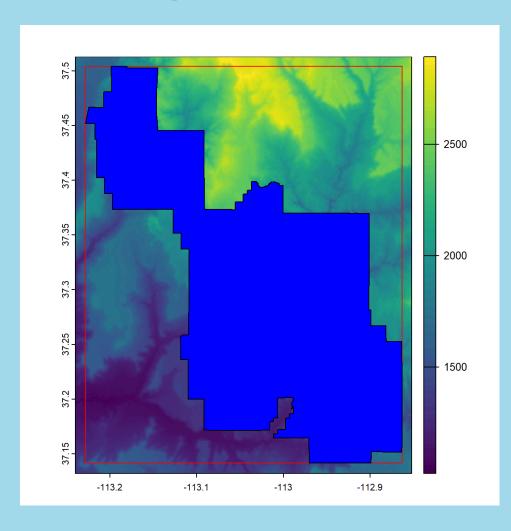
# Modifying the Extent

#### Dealing with Different Extents

- Raster extents often larger than our analysis
- Reducing memory and computational resources
- Making attractive maps



#### Using terra::crop()



- Coordinate Reference System must be the same for both objects
- Crop is based on the (converted)
   SpatExtent of the 2nd object
- snap describes how y will be aligned to the raster
- Returns all data within the extent

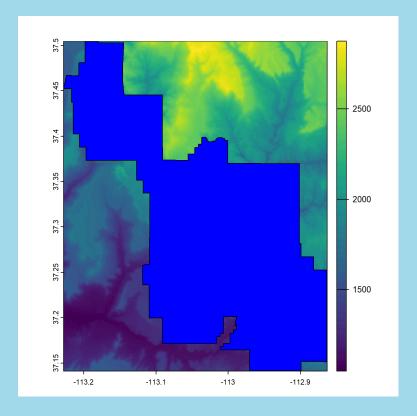
#### Using terra::crop()

```
library(sf)
library(terra)
library(spDataLarge)
srtm = rast(system.file("raster/srtm.tif", package =
  zion = read_sf(system.file("vector/zion.gpkg", package
  zion = st_transform(zion, crs(srtm))

crs(srtm) == crs(zion)
```

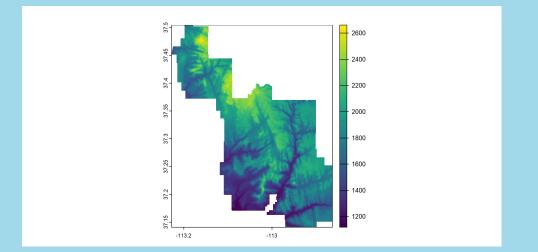
#### [1] TRUE

1 srtm.crop <- crop(x=srtm, y=zion, snap="near")</pre>

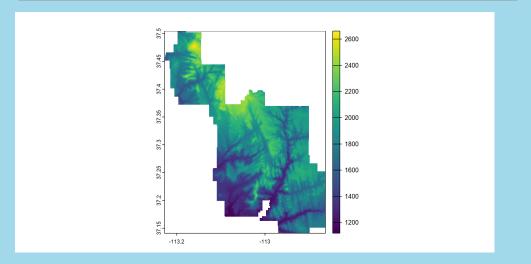


#### Using mask()

- Often want to get rid of all values outside of vector
- Can set mask=TRUE in crop() (y must be SpatVector)
- Or use mask()
  - 1 srtm.crop.msk <- crop(x=s:
  - 2 plot(srtm.crop.msk)



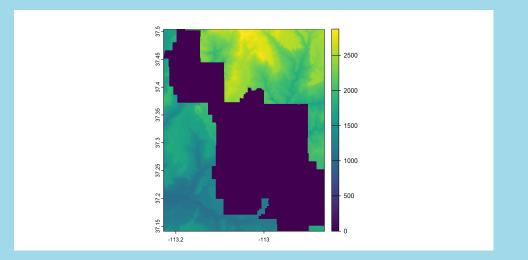
- 1 srtm.msk <- mask(srtm.cro
- 2 plot(srtm.msk)



#### Using mask()

- Allows more control over what the mask does
- Can set maskvalues and updatevalues to change the resulting raster
- Can also use **inverse** to mask out the vector
  - 1 srtm.msk <- mask(srtm.crop
    2 plot(srtm.msk)</pre>
- 2500 2000 1500 1000 500 -1000

- l srtm.msk <- mask(srtm.crop
- 2 plot(srtm.msk)



#### Extending boundaries

- Vector slightly larger than raster
- Especially when using buffered datasets
- Can use extend
- Not exact; depends on snap()

#### **Practice**

**mosaic** is a funciton that combines adjacent rasters, but they need to have the same origin and resolution. Let's practice preparing some rasters for a mosaic.

- 1. Load wildfire hazard data from the rasterexample folder for OR and ID: Copy of CRPS\_OR.tif and Copy of CRPS\_ID.tif.
- 2. These rasters have a fine resolution that will make our calculations slow. Transform them to have a resolution of 900 m.
- 3. Do the rasters have the same CRS, origin, resolution, and extent? Check this with ==.
- 4. Use new functions from this lecture to align the properties mentioned in #3 (plot often to check your work). Why not use project?
- 5. mosaic the two rasters together.