# ESM 263 HW2

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#### Loading and exploring data

## [1] "California" "Cities"

Check what layers are in the basemap:

```
st_layers("HW2/data_hw2/basemap.gpkg")$name
```

"ROI"

"Streets"

"County"

Read them in:

```
california <- read_sf("HW2/data_hw2/basemap.gpkg", layer = "California")

cities <- read_sf("HW2/data_hw2/basemap.gpkg", layer = "Cities")

county <- read_sf("HW2/data_hw2/basemap.gpkg", layer = "County")

ROI <- read_sf("HW2/data_hw2/basemap.gpkg", layer = "ROI")

streets <- read_sf("HW2/data_hw2/basemap.gpkg", layer = "streets")</pre>
```

For this and any other layers we can do a little exploring. For example, we can check the variable names:

```
names(california)
```

```
## [1] "OBJECTID" "NAME_PCASE" "NAME_UCASE" "FMNAME_PC" "FMNAME_UC"

## [6] "ABBREV" "NUM" "ABCODE" "FIPS" "ANSI"

## [11] "ISLAND" "Shape_Leng" "Shape_Length" "Shape_Area" "geom"
```

Note that in an sf object, there's always a column at the end called geom; this contains all the spatial information.

We can also do a quick plot to see what we have in the california shape. Looks like counties.

```
tm_shape(st_geometry(california)) +

tm_polygons()
```



I'm also interested to see the ROI (region of interest). In the html version of this file, you'll be able to see this against a base layer and zoom around etc, as the map view mode is interactive, but in the pdf version, it's a static map. We can see that we're only interested in what looks like the downtown area including the harbor.

For the static map, I use the read\_osm function, where osm stands for Open Street Map, to create a base layer. This requires a bounding box which is represented by the Santa Barbara city feature. This is a multipolygon with two parts, the actual city, and the airport, so I turn it into a polygon with two separate features, and select the second one, Santa Barbara proper.

```
# Create feature for bounding box and read Open Street Map layer
sb <- st_geometry(cities[cities$CITY == "Santa Barbara",])
sb_polygon <- st_cast(sb, to = "POLYGON")
osm <- read_osm(st_buffer(sb_polygon[2], 300), type = "osm") # use buffer to
# have a bit of margin around the map</pre>
```

```
# Map

tm_shape(osm) +

tm_rgb(alpha = 0.7) +

tm_shape(sb_polygon[2]) +

tm_borders(col = "black", lwd = 2) +

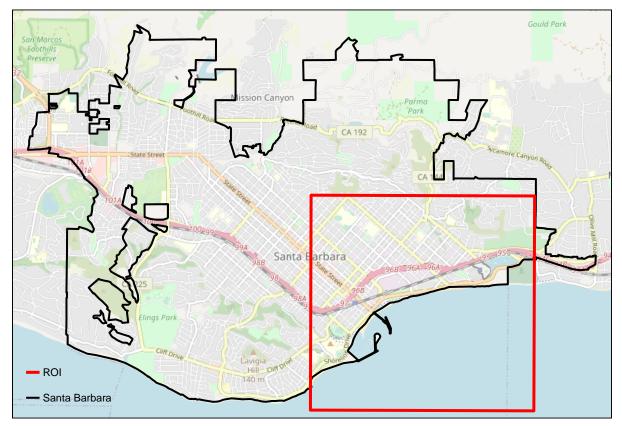
tm_shape(ROI) +

tm_borders(col = "red", lwd = 3) +

tm_add_legend(type = "line", lwd = 3, col = "red", labels = "ROI", alpha = 1) +

tm_add_legend(type = "line", lwd = 2, col = "black", labels = "Santa Barbara", alpha = 1) +

tm_layout(legend.position = c("left", "bottom"))
```

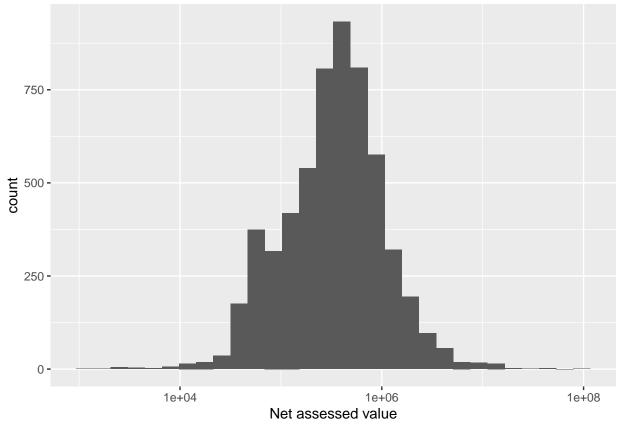


There's only one layer, "parcels", in the parcels file so we'll read that in.

```
parcels <- read_sf("HW2/data_hw2/parcels.gpkg")</pre>
```

The variable we're interested in is NET\_AV, so we can check what that looks like (on a log 10 scale to make it easier to read).

```
ggplot(parcels) +
geom_histogram(aes(x = NET_AV)) +
scale_x_log10() +
xlab("Net assessed value")
```



Seems like most parcels are valued at just under a million dollars, but some are worth tens of millions of dollars.

For the inundation scenarios, I combine all the layers into a single sf object.

```
# Get layer names
inund_layers <- st_layers("HW2/data_hw2/inundation_scenarios.gpkg")$name

# Start with one layer and then row bind the others onto it with a loop
scenarios <- st_read("HW2/data_hw2/inundation_scenarios.gpkg", layer = inund_layers[1], quiet = TRUE)
for(i in 2:length(inund_layers)) {
    scenarios <- rbind(scenarios, read_sf("HW2/data_hw2/inundation_scenarios.gpkg", layer = inund_layers[
}</pre>
```

Let's look at one of these scenarios:

```
# New bounding box for the base layer
osm_ROI <- read_osm(st_buffer(ROI, 500), type = "osm")

# Map

tm_shape(osm_ROI) +
 tm_rgb(alpha = 0.7) +
 tm_shape(st_geometry(filter(scenarios, GRIDCODE == 10))) +
 tm_fill(col = "lightblue", alpha = 0.8)</pre>
```



Seems like this represents current land area that would be inundated under the scenario in question.

## Spatial join

What we want here is the total value of all parcels that fall within the inundated area for each scenario.

```
# Calculate the parcel areas
parcels$area <- st_area(parcels) # calculate area
units(parcels$area) <- with(ud_units, ha) # convert from m^2 to ha</pre>
```

```
parcels$area <- drop_units(parcels$area) # roundabout way of doing this,

# but it is less prone to human error to use the units package for conversions

# Do the join and summarize for each of the three variables

scenarios <- scenarios %>%

st_join(parcels, join = st_intersects) %>%

group_by(GRIDCODE) %>% # this is the ID of each scenario

summarize(parcel_count = n(), net_value = round(sum(NET_AV)/1e6, 0), area = round(sum(area), 0))

# Rename GRIDCODE column

names(scenarios)[1] <- "scenario"</pre>
```

#### Results: table

Now we can turn this into a table.

```
st_drop_geometry(scenarios) %>%
kbl(col.names = c("Sea-level rise (m)", "Parcel count", "Net loss ($m)", "Area flooded (ha)")) %>%
kable_material(c("striped", "hover"))
```

	T.		
Sea-level rise (m)	Parcel count	Net loss (\$m)	Area flooded (ha)
1	60	51	43
2	89	88	108
3	227	195	189
4	620	566	301
5	1275	909	354
6	1863	1215	392
7	2196	1372	436
8	2614	1555	465
9	2958	1712	486
10	3287	1881	512
	-		

# Results: map

In the map, the numbers 1 through 10 represent the number of meters of sea-level rise associated with each scenario and the intensity of the color represents the amount of property value lost.

```
tmap_mode("plot")
tm_shape(scenarios) +
 tm_polygons("net_value", title = "Net loss ($m)") +
 tm_facets(by = "scenario", nrow = 5, ncol = 2) +
 tm_layout(main.title = "Inundation scenarios for downtown Santa Barbara",
            legend.position = c("right", "bottom"),
           main.title.size = 0.8)
```

## Inundation scenarios for downtown Santa Barbara

