

## Homework 8

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This homework is due on April 19, 2021 at 11:00pm. Please submit as a pdf file on Canvas.

In this homework, we will work with two datasets, `US_counties` and `US_census`. The dataset `US_counties` contains the geometry of each county in the US and thus can be used for drawing maps. The dataset `US_census` contains numerous pieces of information about US counties obtained from the US census. Both datasets have a column `FIPS` which can be used to uniquely identify each county in each dataset.

```
# data preparation
US_counties <- readRDS(url("https://wilkelab.org/SDS375/datasets/US_counties.rds")) %>%
  rename(FIPS = GEOID)

# workaround for missing rgdal on edupod
st_crs(US_counties) <- NA

US_census <- read_csv(
  "https://wilkelab.org/SDS375/datasets/US_census.csv",
  col_types = cols(FIPS = "c")
)
```

`US_counties`

```
## Simple feature collection with 3142 features and 11 fields
## geometry type:  MULTIPOLYGON
## dimension:      XY
## bbox:           xmin: -3683715 ymin: -2839538 xmax: 2258154 ymax: 1558935
## CRS:            NA
## First 10 features:
##   STATEFP COUNTYFP COUNTYNS AFFGEOID FIPS NAME LSAD ALAND
## 1      06      075 00277302 05000000US06075 06075 San Francisco 06 121485107
## 2      25      025 00606939 05000000US25025 25025 Suffolk 06 150855462
## 3      31      007 00835826 05000000US31007 31007 Banner 06 1932676697
## 4      37      181 01008591 05000000US37181 37181 Vance 06 653705784
## 5      48      421 01383996 05000000US48421 48421 Sherman 06 2390651189
## 6      50      011 01461762 05000000US50011 50011 Franklin 06 1641633748
## 7      51      680 01498429 05000000US51680 51680 Lynchburg 25 126842531
## 8      05      099 00069169 05000000US05099 05099 Nevada 06 1600192985
## 9      08      014 01945881 05000000US08014 08014 Broomfield 06 85478932
## 10     16      065 00394803 05000000US16065 16065 Madison 06 1215396379
##   AWATER name state_code geometry
## 1 479107241 California CA MULTIPOLYGON (((-2283315 35...
## 2 160479920 Massachusetts MA MULTIPOLYGON (((2009657 799...
## 3 397069 Nebraska NE MULTIPOLYGON (((-664543.3 4...
## 4 42187365 North Carolina NC MULTIPOLYGON (((1544259 321...
## 5 428754 Texas TX MULTIPOLYGON (((-546533.1 -...
## 6 150930318 Vermont VT MULTIPOLYGON (((1780210 102...
## 7 1428787 Virginia VA MULTIPOLYGON (((1458790 122...
## 8 7254365 Arkansas AR MULTIPOLYGON (((232307.6 -4...
```

```
## 9 1411395 Colorado CO MULTIPOLYGON (((-775389.1 3...
## 10 10500827 Idaho ID MULTIPOLYGON (((-1272146 82...
```

```
US_census
```

```
## # A tibble: 3,143 x 53
##   state name FIPS pop2010 pop2000 age_under_5 age_under_18 age_over_65 female
##   <chr> <chr> <chr>   <dbl>   <dbl>   <dbl>       <dbl>       <dbl> <dbl>
## 1 Alab~ Auta~ 01001   54571   43671     6.6        26.8        12    51.3
## 2 Alab~ Bald~ 01003  182265  140415     6.1         23        16.8   51.1
## 3 Alab~ Barb~ 01005   27457   29038     6.2        21.9        14.2   46.9
## 4 Alab~ Bibb~ 01007   22915   20826     6         22.7        12.7   46.3
## 5 Alab~ Blou~ 01009   57322   51024     6.3        24.6        14.7   50.5
## 6 Alab~ Bull~ 01011   10914   11714     6.8        22.3        13.5   45.8
## 7 Alab~ Butl~ 01013   20947   21399     6.5        24.1        16.7    53
## 8 Alab~ Calh~ 01015  118572  112249     6.1        22.9        14.3   51.8
## 9 Alab~ Cham~ 01017   34215   36583     5.7        22.5        16.7   52.2
## 10 Alab~ Cher~ 01019   25989   23988     5.3        21.4        17.9   50.4
## # ... with 3,133 more rows, and 44 more variables: white <dbl>, black <dbl>,
## #   native <dbl>, asian <dbl>, pac_isl <dbl>, two_plus_races <dbl>,
## #   hispanic <dbl>, white_not_hispanic <dbl>, no_move_in_one_plus_year <dbl>,
## #   foreign_born <dbl>, foreign_spoken_at_home <dbl>, hs_grad <dbl>,
## #   bachelors <dbl>, veterans <dbl>, mean_work_travel <dbl>,
## #   housing_units <dbl>, home_ownership <dbl>, housing_multi_unit <dbl>,
## #   median_val_owner_occupied <dbl>, households <dbl>,
## #   persons_per_household <dbl>, per_capita_income <dbl>,
## #   median_household_income <dbl>, poverty <dbl>,
## #   private_nonfarm_establishments <dbl>, private_nonfarm_employment <dbl>,
## #   percent_change_private_nonfarm_employment <dbl>,
## #   nonemployment_establishments <dbl>, firms <dbl>, black_owned_firms <dbl>,
## #   native_owned_firms <dbl>, asian_owned_firms <dbl>,
## #   pac_isl_owned_firms <dbl>, hispanic_owned_firms <dbl>,
## #   women_owned_firms <dbl>, manufacturer_shipments_2007 <dbl>,
## #   mercent_whole_sales_2007 <dbl>, sales <dbl>, sales_per_capita <dbl>,
## #   accommodation_food_service <dbl>, building_permits <dbl>,
## #   fed_spending <dbl>, area <dbl>, density <dbl>
```

**Problem 1: (6 pts)** Make a choropleth map of the percent home-ownership (column `home_ownership` in `US_census`) for all counties in the US. Choose an appropriate color scale and design for this plot. You may notice that there is one county in Alaska for which home-ownership data is not available. Write data analysis code to identify this county.

#### Hints:

1. Use `theme_void()` as your theme
2. You will have to join `US_counties` and `US_census`. Join them by the `FIPS` column.
3. To make nice percent labels, you can use `label = scales::label_percent(scale = 1)` in your color scale function.
4. To find rows with missing data, you may want to use the function `is.na()`.

Grade breakdown: 2pt for the plot, 2pt for the plot design, and 2pt for identifying the county in Alaska for which home ownership data is not available.

```
# your code goes here
US_data <- US_counties %>% left_join(US_census, by = "FIPS")
```

# US\_data

```
## Simple feature collection with 3142 features and 63 fields
## geometry type:  MULTIPOLYGON
## dimension:      XY
## bbox:           xmin: -3683715 ymin: -2839538 xmax: 2258154 ymax: 1558935
## CRS:            NA
## First 10 features:
```

##	STATEFP	COUNTYFP	COUNTYNS	AFFGEOID	FIPS	NAME	LSAD	ALAND
## 1	06	075	00277302	0500000US06075	06075	San Francisco	06	121485107
## 2	25	025	00606939	0500000US25025	25025	Suffolk	06	150855462
## 3	31	007	00835826	0500000US31007	31007	Banner	06	1932676697
## 4	37	181	01008591	0500000US37181	37181	Vance	06	653705784
## 5	48	421	01383996	0500000US48421	48421	Sherman	06	2390651189
## 6	50	011	01461762	0500000US50011	50011	Franklin	06	1641633748
## 7	51	680	01498429	0500000US51680	51680	Lynchburg	25	126842531
## 8	05	099	00069169	0500000US05099	05099	Nevada	06	1600192985
## 9	08	014	01945881	0500000US08014	08014	Broomfield	06	85478932
## 10	16	065	00394803	0500000US16065	16065	Madison	06	1215396379

```
##
```

##	AWATER	name.x	state_code	state	name.y
## 1	479107241	California	CA	California	San Francisco County
## 2	160479920	Massachusetts	MA	Massachusetts	Suffolk County
## 3	397069	Nebraska	NE	Nebraska	Banner County
## 4	42187365	North Carolina	NC	North Carolina	Vance County
## 5	428754	Texas	TX	Texas	Sherman County
## 6	150930318	Vermont	VT	Vermont	Franklin County
## 7	1428787	Virginia	VA	Virginia	Lynchburg city
## 8	7254365	Arkansas	AR	Arkansas	Nevada County
## 9	1411395	Colorado	CO	Colorado	Broomfield County
## 10	10500827	Idaho	ID	Idaho	Madison County

```
##
```

##	pop2010	pop2000	age_under_5	age_under_18	age_over_65	female	white	black
## 1	805235	776733	4.4	13.4	13.6	49.3	48.5	6.1
## 2	722023	689807	5.5	17.5	10.5	51.8	56.0	21.6
## 3	690	819	5.2	22.3	19.7	48.7	95.7	0.0
## 4	45422	42954	6.8	25.4	14.1	53.1	44.2	49.9
## 5	3034	3186	7.1	30.2	13.3	49.0	88.3	0.5
## 6	47746	45417	6.3	24.7	12.2	50.4	95.6	0.4
## 7	75568	65269	6.1	19.6	14.0	53.1	64.4	29.3
## 8	8997	9955	6.6	23.7	17.7	51.0	65.9	30.7
## 9	55889	NA	7.1	26.2	9.9	50.4	86.1	1.1
## 10	37536	27467	9.6	26.0	5.6	51.6	93.9	0.5

```
##
```

##	native	asian	pac_isl	two_plus_races	hispanic	white_not_hispanic
## 1	0.5	33.3	0.4	4.7	15.1	41.9
## 2	0.4	8.2	0.0	3.9	19.9	48.1
## 3	0.4	0.0	0.0	0.7	3.8	95.1
## 4	0.3	0.4	NA	1.3	6.7	42.1
## 5	0.8	0.2	0.0	1.5	40.4	58.1
## 6	1.0	0.5	NA	2.1	1.2	94.9
## 7	0.3	2.5	0.0	2.2	3.0	63.0
## 8	0.3	0.3	0.0	1.4	2.4	65.1
## 9	0.6	6.1	NA	2.8	11.1	79.4
## 10	0.3	0.9	0.1	1.5	5.9	91.2

```
##
```

##	no_move_in_one_plus_year	foreign_born	foreign_spoken_at_home	hs_grad
----	--------------------------	--------------	------------------------	---------

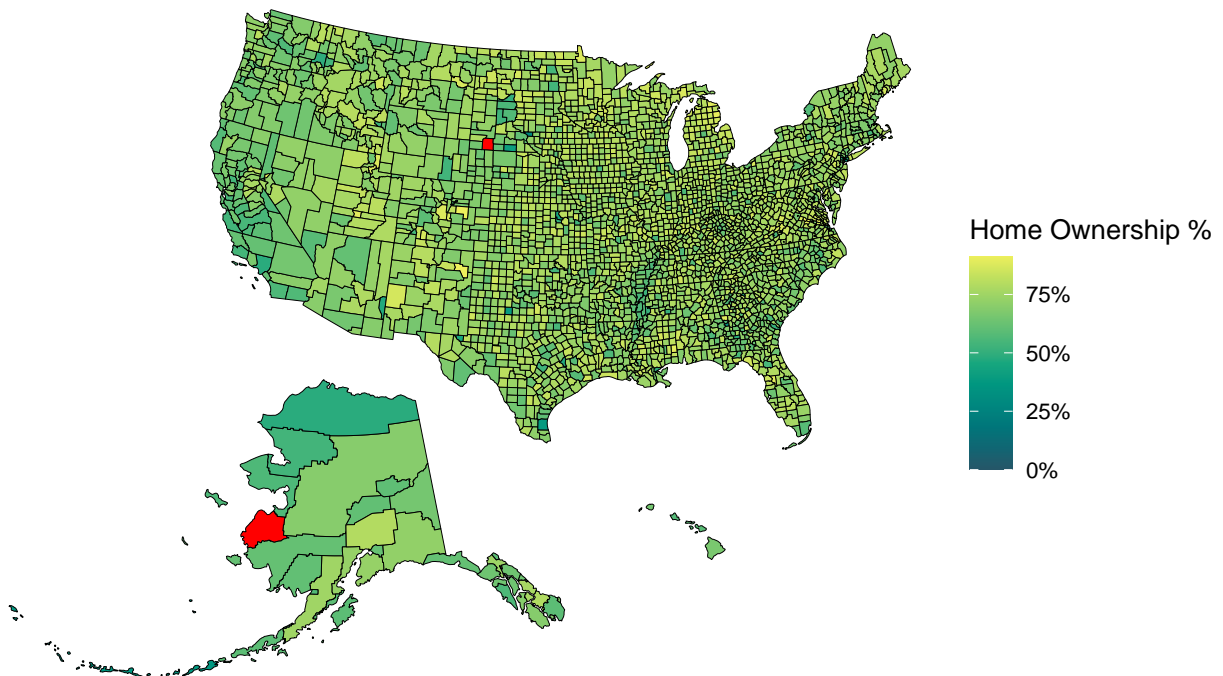
## 1		84.0	35.6	45.4	85.7
## 2		76.7	27.7	36.8	83.1
## 3		91.3	3.8	3.3	94.1
## 4		86.9	3.9	6.7	72.6
## 5		83.9	17.9	31.9	74.4
## 6		88.1	3.2	4.8	87.8
## 7		76.9	4.2	6.3	83.1
## 8		84.7	0.4	1.0	78.3
## 9		82.8	8.0	12.4	95.3
## 10		58.1	3.0	9.7	95.2
##	bachelors veterans	mean_work_travel	housing_units	home_ownership	
## 1	51.2	33078	29.4	376942	37.5
## 2	38.9	25698	28.5	315522	36.5
## 3	25.1	49	25.6	369	66.3
## 4	10.7	3272	23.1	20082	65.2
## 5	19.1	233	16.4	1252	78.7
## 6	21.2	4054	26.6	21588	74.8
## 7	28.3	6213	16.3	31992	56.3
## 8	10.7	758	23.8	4563	71.3
## 9	44.4	4257	26.8	22646	74.4
## 10	30.0	1079	15.2	11280	52.0
##	housing_multi_unit	median_val_owner_occupied	households		
## 1		66.6	785200	335956	
## 2		80.7	384500	283954	
## 3		7.2	76900	320	
## 4		9.6	98100	16473	
## 5		4.0	80800	971	
## 6		16.9	202800	18482	
## 7		29.8	139100	27875	
## 8		5.9	62300	3782	
## 9		21.7	270500	20116	
## 10		42.2	169700	9868	
##	persons_per_household	per_capita_income	median_household_income	poverty	
## 1		2.31	45478	71304	11.9
## 2		2.33	30720	50597	20.6
## 3		2.25	22042	34063	16.1
## 4		2.67	17622	34025	27.5
## 5		3.09	21587	50069	12.9
## 6		2.54	24767	53623	10.5
## 7		2.35	21586	37058	22.7
## 8		2.38	21020	38375	20.0
## 9		2.63	35836	75590	5.5
## 10		3.49	13735	35461	32.2
##	private_nonfarm_establishments	private_nonfarm_employment			
## 1		30490	492689		
## 2		19210	562929		
## 3		6	0		
## 4		878	13177		
## 5		56	263		
## 6		1028	11745		
## 7		2255	54809		
## 8		134	1842		
## 9		1677	29760		
## 10		780	14136		

##	percent_change_private_nonfarm_employment	nonemployment_establishments	
## 1	-11.3	78347	
## 2	-1.4	41757	
## 3	NA	48	
## 4	-10.4	2198	
## 5	-56.9	217	
## 6	0.9	3521	
## 7	-4.7	3556	
## 8	-6.3	423	
## 9	NA	4181	
## 10	15.5	2195	
##	firms	black_owned_firms	native_owned_firms
## 1	105030	2.7	NA
## 2	55613	11.0	0.7
## 3	39	NA	NA
## 4	3089	25.9	NA
## 5	138	NA	NA
## 6	4754	NA	NA
## 7	4794	11.6	NA
## 8	622	NA	NA
## 9	6284	1.3	NA
## 10	2942	NA	NA
##	pac_isl_owned_firms	hispanic_owned_firms	women_owned_firms
## 1	0.2	6.6	30.1
## 2	0.1	8.3	29.5
## 3	NA	NA	NA
## 4	NA	NA	28.5
## 5	NA	NA	NA
## 6	NA	NA	23.6
## 7	NA	NA	26.4
## 8	NA	NA	NA
## 9	NA	2.0	NA
## 10	NA	NA	15.2
##	manufacturer_shipments_2007	mercent_whole_sales_2007	sales
## 1	2077457	10562176	12399960
## 2	3703400	8914822	7565785
## 3	0	0	0
## 4	639233	NA	582116
## 5	0	NA	10079
## 6	NA	413283	545346
## 7	2801602	541451	1665130
## 8	NA	NA	136792
## 9	2255472	NA	1139808
## 10	NA	161967	407686
##	sales_per_capita	accommodation_food_service	building_permits
## 1	15516	5039171	779
## 2	10381	3795879	480
## 3	0	0	0
## 4	13553	53898	83
## 5	3486	NA	8
## 6	11383	39606	161
## 7	23171	189961	144
## 8	14660	3228	1
## 9	21240	163445	232
			fed_spending
## 1			10707798
## 2			16051877
## 3			8130
## 4			408868
## 5			39026
## 6			415662
## 7			1264105
## 8			98360
## 9			442693

```
## 10      10971      32041      249      121324
##      area density      geometry
## 1  46.87 17179.1 MULTIPOLYGON (((-2283315 35...
## 2  58.15 12415.7 MULTIPOLYGON (((2009657 799...
## 3  746.11    0.9 MULTIPOLYGON (((-664543.3 4...
## 4  253.52   179.2 MULTIPOLYGON (((1544259 321...
## 5  923.04    3.3 MULTIPOLYGON (((-546533.1 -...
## 6  633.71   75.3 MULTIPOLYGON (((1780210 102...
## 7   49.13  1538.2 MULTIPOLYGON (((1458790 122...
## 8  617.84   14.6 MULTIPOLYGON (((232307.6 -4...
## 9   33.03  1691.9 MULTIPOLYGON (((-775389.1 3...
## 10 469.21   80.0 MULTIPOLYGON (((-1272146 82...
```

```
ggplot(US_data) +
  geom_sf(
    aes(fill= home_ownership),
    color = "black", size = 0.1
  ) +
  scale_fill_continuous_sequential(palette = "ag_GrnYl",
    rev = FALSE,
    na.value = "Red",
    label = scales::label_percent(scale = 1),
    name = "Home Ownership %") +

  coord_sf() +
  theme_void()
```



```
US_missing <- US_data %>%
  filter(name.x == "Alaska") %>%
  filter(is.na(home_ownership))
```

```
US_missing$NAME
```

```
## [1] Kusilvak
```

```
## 1910 Levels: Abbeville Acadia Accomack Ada Adair Adams Addison ... Ziebach
```

```
##the county in Alaska for which home ownership data is not available: Kusilvak
```

**Problem 2: (4 pts)** Make a choropleth map of the percent foreign born (column `foreign_born` in `US_census`) for the counties in Texas only. Use a different color scale than you used for Problem 1 and use a theme that shows longitude and latitude (nearly any theme other than `theme_void()` will work).

Grade breakdown: 3pt for the plot, 1pt for the plot design

```
# your code goes here
```

```
US_data_fb_TX <- US_counties %>%
```

```
  left_join(US_census, by = "FIPS") %>%
```

```
  filter(state == "Texas")
```

```
US_data_fb_TX
```

```
## Simple feature collection with 254 features and 63 fields
```

```
## geometry type:  MULTIPOLYGON
```

```
## dimension:      XY
```

```
## bbox:           xmin: -999736.2 ymin: -1295598 xmax: 235280.5 ymax: -88854.17
```

```
## CRS:            NA
```

```
## First 10 features:
```

##	STATEFP	COUNTYFP	COUNTYNS	AFFGEOID	FIPS	NAME	LSAD	ALAND
## 1	48	421	01383996	0500000US48421	48421	Sherman	06	2390651189
## 2	48	493	01384032	0500000US48493	48493	Wilson	06	2081662847
## 3	48	115	01383843	0500000US48115	48115	Dawson	06	2331781556
## 4	48	069	01383820	0500000US48069	48069	Castro	06	2316573415
## 5	48	279	01383926	0500000US48279	48279	Lamb	06	2631897577
## 6	48	385	01383978	0500000US48385	48385	Real	06	1810812906
## 7	48	359	01383965	0500000US48359	48359	Oldham	06	3886337121
## 8	48	127	01383849	0500000US48127	48127	Dimmit	06	3441795365
## 9	48	305	01383938	0500000US48305	48305	Lynn	06	2309924845
## 10	48	171	01383871	0500000US48171	48171	Gillespie	06	2740716834

##	AWATER	name.x	state_code	state	name.y	pop2010	pop2000
----	--------	--------	------------	-------	--------	---------	---------

## 1	428754	Texas	TX	Texas	Sherman County	3034	3186
------	--------	-------	----	-------	----------------	------	------

## 2	12111367	Texas	TX	Texas	Wilson County	42918	32408
------	----------	-------	----	-------	---------------	-------	-------

## 3	4720730	Texas	TX	Texas	Dawson County	13833	14985
------	---------	-------	----	-------	---------------	-------	-------

## 4	12584396	Texas	TX	Texas	Castro County	8062	8285
------	----------	-------	----	-------	---------------	------	------

## 5	3973764	Texas	TX	Texas	Lamb County	13977	14709
------	---------	-------	----	-------	-------------	-------	-------

## 6	2346258	Texas	TX	Texas	Real County	3309	3047
------	---------	-------	----	-------	-------------	------	------

## 7	2339716	Texas	TX	Texas	Oldham County	2052	2185
------	---------	-------	----	-------	---------------	------	------

## 8	14576538	Texas	TX	Texas	Dimmit County	9996	10248
------	----------	-------	----	-------	---------------	------	-------

## 9	4169719	Texas	TX	Texas	Lynn County	5915	6550
------	---------	-------	----	-------	-------------	------	------

## 10	9015052	Texas	TX	Texas	Gillespie County	24837	20814
-------	---------	-------	----	-------	------------------	-------	-------

##	age_under_5	age_under_18	age_over_65	female	white	black	native	asian	pac_isl
----	-------------	--------------	-------------	--------	-------	-------	--------	-------	---------

## 1	7.1	30.2	13.3	49.0	88.3	0.5	0.8	0.2	0.0
------	-----	------	------	------	------	-----	-----	-----	-----

## 2	6.2	26.4	12.6	50.2	89.1	1.6	0.7	0.4	NA
------	-----	------	------	------	------	-----	-----	-----	----

## 3	7.6	24.7	14.0	43.5	76.7	6.5	0.7	0.4	0.1
------	-----	------	------	------	------	-----	-----	-----	-----

## 4	8.7	31.3	13.4	49.3	68.6	2.0	0.9	NA	0.0
------	-----	------	------	------	------	-----	-----	----	-----

## 5	8.1	29.3	15.9	50.2	74.4	4.3	0.9	0.1	0.0
------	-----	------	------	------	------	-----	-----	-----	-----

## 6	4.9	19.9	25.8	50.5	90.5	0.7	1.1	0.1	NA
------	-----	------	------	------	------	-----	-----	-----	----

## 7	4.9	34.0	12.4	49.6	90.2	3.0	0.6	0.8	NA
## 8	8.2	30.0	14.2	51.6	88.6	1.0	0.3	0.6	0.0
## 9	7.4	27.8	15.8	50.2	80.3	2.1	1.2	0.1	NA
## 10	5.0	20.3	26.5	51.8	91.3	0.4	0.6	0.4	NA
##	two_plus_races hispanic white_not_hispanic no_move_in_one_plus_year								
## 1	1.5	40.4		58.1				83.9	
## 2	2.1	38.2		58.7				87.7	
## 3	2.3	53.4		39.1				85.9	
## 4	1.6	59.9		37.3				84.8	
## 5	2.3	51.7		43.1				83.1	
## 6	2.0	24.6		72.5				87.7	
## 7	1.5	11.8		82.8				85.5	
## 8	1.6	86.2		12.2				82.8	
## 9	2.3	46.4		50.4				84.7	
## 10	1.2	20.0		78.4				89.0	
##	foreign_born foreign_spoken_at_home hs_grad bachelors veterans								
## 1	17.9		31.9	74.4	19.1			233	
## 2	4.8		26.1	84.5	18.2			4458	
## 3	6.9		37.8	67.0	8.7			633	
## 4	18.2		49.0	68.1	14.9			408	
## 5	10.4		37.4	71.9	13.8			813	
## 6	5.5		21.0	77.6	19.4			358	
## 7	2.8		8.7	82.3	29.8			127	
## 8	9.6		69.9	61.0	12.6			469	
## 9	6.6		34.5	74.2	15.9			312	
## 10	6.2		23.5	86.4	26.8			2997	
##	mean_work_travel housing_units home_ownership housing_multi_unit								
## 1	16.4	1252		78.7				4.0	
## 2	32.5	16766		84.5				3.7	
## 3	12.2	5220		75.7				6.6	
## 4	18.1	3166		77.0				5.5	
## 5	16.2	6128		75.1				6.1	
## 6	20.2	2599		75.5				0.0	
## 7	16.6	841		71.0				9.6	
## 8	22.7	4350		69.5				10.6	
## 9	16.8	2676		73.9				4.8	
## 10	20.0	12778		75.5				7.0	
##	median_val_owner_occupied households persons_per_household per_capita_income								
## 1		80800	971			3.09		21587	
## 2		124100	14455			2.82		25149	
## 3		48000	4424			2.58		15288	
## 4		68300	2677			2.90		16073	
## 5		50100	4804			2.90		17553	
## 6		75300	1335			2.39		15074	
## 7		86900	710			2.72		22504	
## 8		49000	3600			2.67		14045	
## 9		66900	2165			2.77		19752	
## 10		191100	10413			2.27		28072	
##	median_household_income poverty private_nonfarm_establishments								
## 1		50069	12.9					56	
## 2		60493	9.0					500	
## 3		33623	19.1					285	
## 4		35087	23.3					179	
## 5		35458	17.9					274	



## 6	29186	26.8		84
## 7	51111	13.4		34
## 8	25882	36.4		164
## 9	43672	16.9		95
## 10	52682	8.0		890
##	private_nonfarm_employment	percent_change_private_nonfarm_employment		
## 1	263			-56.9
## 2	4647			49.0
## 3	2743			7.7
## 4	1093			-24.2
## 5	2921			0.5
## 6	539			26.2
## 7	0			NA
## 8	1780			23.5
## 9	704			-10.1
## 10	8061			29.3
##	nonemployment_establishments	firms	black_owned_firms	native_owned_firms
## 1	217	138	NA	NA
## 2	2813	3222	NA	NA
## 3	645	887	NA	NA
## 4	474	534	NA	NA
## 5	708	1038	NA	NA
## 6	437	NA	NA	NA
## 7	184	114	NA	NA
## 8	798	1017	NA	NA
## 9	307	NA	NA	NA
## 10	3138	4379	NA	NA
##	asian_owned_firms	pac_isl_owned_firms	hispanic_owned_firms	women_owned_firms
## 1	NA	NA	NA	NA
## 2	NA	NA	22.6	NA
## 3	NA	NA	NA	NA
## 4	NA	NA	8.1	NA
## 5	NA	NA	NA	15.9
## 6	NA	NA	NA	NA
## 7	NA	NA	NA	NA
## 8	NA	NA	NA	NA
## 9	NA	NA	NA	NA
## 10	NA	NA	NA	28.7
##	manufacturer_shipments_2007	mercent_whole_sales_2007	sales	sales_per_capita
## 1	0	NA	10079	3486
## 2	0	48987	180452	4609
## 3	0	91324	148370	10747
## 4	0	154599	40007	5579
## 5	NA	84216	95775	6933
## 6	0	NA	9440	3128
## 7	0	NA	16797	8118
## 8	0	NA	74775	7658
## 9	0	NA	19743	3415
## 10	88259		100425	311100
##	accommodation_food_service	building_permits	fed_spending	area_density
## 1	NA	8	39026	923.04
## 2	20847	61	220705	803.73
## 3	12466	1	143793	900.31
## 4	2366	0	76493	894.43

```
## 5          6963          0      137879 1016.18    13.8
## 6          3662          0       34403  699.20     4.7
## 7          2919          4       20356 1500.53     1.4
## 8          8259          5      119429 1328.88     7.5
## 9           795          1       85142  891.87     6.6
## 10         52219         34      182308 1058.21    23.5
##              geometry
## 1 MULTIPOLYGON (((-546533.1 -...
## 2 MULTIPOLYGON (((-234487.5 -...
## 3 MULTIPOLYGON (((-576468.3 -...
## 4 MULTIPOLYGON (((-592119.6 -...
## 5 MULTIPOLYGON (((-603646.3 -...
## 6 MULTIPOLYGON (((-389104.7 -...
## 7 MULTIPOLYGON (((-631641.1 -...
## 8 MULTIPOLYGON (((-402909.8 -...
## 9 MULTIPOLYGON (((-561102.7 -...
## 10 MULTIPOLYGON (((-316455 -77...
```

```
ggplot(US_data_fb_TX) +
  geom_sf(aes(fill = foreign_born),
    color = "black", size = 0.1) +
  scale_fill_continuous_sequential(palette = "Blues",
    label = scales::label_percent(scale = 1),
    name = "Foreign Born %") +
  coord_sf() +
  theme_bw() +
  theme(axis.text.x = element_text(size = 7))
```

