## Census data

We load previously loaded census data. The code for fetching this data is also shown in this section.

```
load(here("API KEY.rda"))
options(cancensus.api_key = api_key)
options(cancensus.cache_path = "cache")
vectors <- c("v_CA21_1", "v_CA21_6", "v_CA21_452", "v_CA21_449", "v_CA21_1040", "v_CA21_1085", "v_CA21_
region_DA <- c("59154012", "59154105", "59154090", "59150936", "59154101", "59154104",
               "59154035", "59154103", "59154102", "59154034", "59150945", "59154091",
               "59154093", "59154099", "59150946", "59154100", "59154078", "59154079",
               "59154082", "59154081", "59154080", "59150939", "59150938", "59154083",
               "59154095", "59154084", "59150941", "59150942", "59154085", "59154088",
               "59154087", "59154089", "59154097", "59154098", "59154096", "59154092",
               "59154013", "59150952")
census_data <- get_census(</pre>
 dataset = "CA21",
 regions = list(CSD = "5915022", DA = region_DA),
 vectors = vectors,
 labels = "detailed",
  geo_format = "sf",
 level = "DA"
census_data <- census_data %>%
 mutate(pop_density = `v_CA21_1: Population, 2021` / `Shape Area`)
can api key <- ""
save(census_data, file = "../../data/census.rda")
load(here("data/census.rda"))
n <- nrow(census_data)</pre>
kable(head(census_data[, 1:5]), format = "latex", booktabs = TRUE, caption = "Census Data, 2021 (trunca
 kable_styling(latex_options = c("striped", "hold_position"))
```

Table 1: Census Data, 2021 (truncated rows and columns)

Shape Area	Type	Households	Quality Flags	name	geometry
0.2991	DA	266	0	59150307	MULTIPOLYGON (((-123.0231 4
0.1096	DA	218	0	59150308	MULTIPOLYGON (((-123.0234 4
0.1119	DA	282	0	59150309	MULTIPOLYGON (((-123.0283 4
0.1094	DA	389	0		MULTIPOLYGON (((-123.0234 4
0.0809	DA	187	0	59150311	MULTIPOLYGON (((-123.0257 4
0.0871	DA	201	0	59150312	MULTIPOLYGON (((-123.0234 4

## Food data

```
food data <- st read(here("data/free-and-low-cost-food-programs.shp")) %>%
     select(
          "program_nam",
          "program_sta",
         "meal_cost",
         "local_areas",
         "latitude",
         "longitude",
         "geometry"
     ) %>%
     drop_na("latitude", "longitude") %>%
     # set to wgs 84 as per can census
     st_set_crs(4326)
## Reading layer `free-and-low-cost-food-programs' from data source
            '/Users/sid/Documents/ubc classes/2024w1/econ 326/foodprograms-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-and-low-cost-food-programs-326/data/free-a
##
            using driver `ESRI Shapefile'
## replacing null geometries with empty geometries
## Simple feature collection with 83 features and 25 fields (with 2 geometries empty)
## Geometry type: POINT
## Dimension:
                                          XY
## Bounding box: xmin: -123.1821 ymin: 49.20725 xmax: -123.0287 ymax: 49.286
## CRS:
                                           NA
# Food data processing
food_count <- food_data %>%
     st_set_geometry(NULL) %>%
     group_by(local_areas) %>%
     summarise(count = n(), .groups = "drop")
food_data_count <- food_data %>%
     left_join(food_count, by = "local_areas") %>%
     distinct(local_areas, .keep_all = TRUE) # one row per neighbourhood
combo_food_census <- census_data %>%
     st_join(food_data_count)
census_data_food <- combo_food_census %>%
     mutate(program_count = replace_na(count, 0),
                     food_density = program_count / `Shape Area`)
head(census_data_food[, 1:5]) %>%
     kable(format = "latex", booktabs = TRUE, caption = "Food Data merged with Census Data, 2021 (truncate
    kable_styling(latex_options = c("striped", "hold_position"))
Crime data
```

Table 2: Food Data merged with Census Data, 2021 (truncated rows and columns)

Shape Area	Type	Households	Quality Flags	name	geometry
0.2991	DA	266	0	59150307	MULTIPOLYGON (((-123.0231 4
0.1096	DA	218	0	59150308	MULTIPOLYGON (((-123.0234 4
0.1119	DA	282	0	59150309	MULTIPOLYGON (((-123.0283 4
0.1094	DA	389	0	59150310	MULTIPOLYGON (((-123.0234 4
0.0809	DA	187	0	59150311	MULTIPOLYGON (((-123.0257 4
0.0871	DA	201	0	59150312	MULTIPOLYGON (((-123.0234 4

```
filter(!is.na(X) & !is.na(Y))

crime_data <- st_as_sf(crime, coords = c("X", "Y"), crs = "+proj=utm +zone=10") %>%
    st_transform(crs = "+proj=longlat +datum=WGS84")

intersections <- st_is_within_distance(census_data, crime_data, sparse = FALSE, dist = 5)

crimes_contained <- rowSums(intersections, dims = 1)

census_data_crime <- census_data %>%
    cbind(crimes_contained) %>%
    mutate(crime_density = crimes_contained / Shape.Area)

unique_crimes <- unique(crime_data$TYPE)

for (type in unique_crimes) {
    type_data <- crime_data %>% filter(TYPE == type)
    intersections <- st_is_within_distance(census_data, type_data, sparse = FALSE, dist = 5)
    sum <- rowSums(intersections, dims = 1)
    df <- as.data.frame(sum)
    census_data_crime <- census_data_crime %>% cbind(df$sum) %>% rename_with(- paste0("crimes_", type), d
}
```

# Training data

Now we merge all of our data-sets into one table so we can feed it into our model.

```
training_data[is.na(training_data)] <- 0</pre>
```

# Model

# Specification

```
# Specification models
# 1. Model with all variables (including interaction term)
reg_all_vars <- lm(crime_density ~</pre>
                   food_density:low_income +
                    food_density +
                   pop_density.x +
                    v_CA21_452 +
                   low_income,
                 data = st_set_geometry(training_data, NULL))
# 2. Model without the interaction term
reg_no_interaction <- lm(crime_density ~</pre>
                          food_density +
                          pop_density.x +
                          v_CA21_452 +
                          low_income,
                        data = st_set_geometry(training_data, NULL))
# 3. Model with only food density, crime density, and low income
reg_food_crime_low_income <- lm(crime_density ~</pre>
                                 food_density +
                                 low income,
                               data = st_set_geometry(training_data, NULL))
# 4. Model with all variables (including interaction term) but replacing v_CA21_452 with v_CA21_449
reg_all_vars_449 <- lm(crime_density ~
                        food_density:low_income +
                        food density +
                        pop_density.x +
                        v_CA21_449 + \#Replaced v_CA21_452 with v_CA21_449
                        low_income,
                      data = st_set_geometry(training_data, NULL))
models_spec <- list(</pre>
  "All Variables" = reg_all_vars,
 "Without Interaction" = reg_no_interaction,
  "Food Density, Crime, Low Income" = reg_food_crime_low_income
# Summary for each specification tested
summary_all_vars <- summary(reg_all_vars)</pre>
summary_no_interaction <- summary(reg_no_interaction)</pre>
summary_food_crime_low_income <- summary(reg_food_crime_low_income)</pre>
```

```
summary_all_vars_449 <- summary(reg_all_vars_449)

# Add to the list of model summaries
model_summaries <- list(
    "All Variables" = summary_all_vars,
    "Without Interaction" = summary_no_interaction,
    "Food Density, Crime, Low Income" = summary_food_crime_low_income,
    "All Variables (with v_CA21_449)" = summary_all_vars_449
)</pre>
```

## RESULTS FROM SPEC

```
stargazer(
  reg_all_vars,
  reg_no_interaction,
  reg food crime low income,
  type = "latex",
  covariate.labels = c(
   "(Intercept)",
   "Food Program Density: Low Income",
   "Food Program Density",
   "Population Density",
   "Average Household Size (v\\CA21\\L452)",
   "Low Income"
  ),
  omit = c("pop_density.x", "v_CA21_452"),
  add.lines = list(
   c("Includes Interaction?", "Yes", "No", "No"),
   c("Population Density Included?", "Yes", "Yes", "No"),
   c("Average Household Size (v\\_CA21\\_452) Included?", "Yes", "Yes", "No")
 ),
 title = "Regression Specification Table",
  digits = 3
)
```

% Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac at gmail.com % Date and time: Sun, Dec 01, 2024 - 13:27:10

#### Model 1:

Table 3: Regression Specification Table

		$Dependent\ variable:$	
		crime_density	
	(1)	(2)	
(Intercept)	-0.203	8.053***	9.6
· · ·	(2.154)	(0.771)	(0
Food Program Density: Low Income	244.697***	252.689***	577.
Ç V	(24.734)	(24.682)	(27
Food Program Density	9.460***		
V V	(2.305)		
Population Density	1,291.286***	1,283.891***	383.
	(62.011)	(62.047)	(18
Includes Interaction?	Yes	No	
Population Density Included?	Yes	Yes	
Average Household Size (v_CA21_452) Included?	Yes	Yes	
Observations	7,860	7,860	7
$\mathbb{R}^2$	0.315	0.313	0
Adjusted $R^2$	0.314	0.313	C
Residual Std. Error	1,050.245  (df = 7854)	1,051.304 (df = 7855)	1,225.029
F Statistic	$720.960^{***} \text{ (df} = 5; 7854)$	$895.186^{***} (df = 4; 7855)$	282.621***

Note: \*p<0.1; \*\*p<0

```
c("Average Household Size (v\\_CA21\\_452) Included?", "Yes", "Yes", "No")
),
title = "Spec 1: All Variables",
digits = 3)
```

% Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac at gmail.com % Date and time: Sun, Dec 01, 2024 - 13:27:10

Table 4: Spec 1: All Variables

	Dependent variable:
	crime_density
(Intercept)	-0.203
	(2.154)
Food Program Density: Low Income	0.043***
	(0.001)
Food Program Density	-517.756***
·	(22.439)
Population Density	244.697***
-	(24.734)
Average Household Size (v_CA21_452)	9.460***
,	(2.305)
Low Income	1,291.286***
	(62.011)
Includes Interaction?	Yes
Population Density Included?	Yes
Average Household Size (v_CA21_452) Included?	Yes
Observations	7,860
$\mathbb{R}^2$	0.315
Adjusted $\mathbb{R}^2$	0.314
Residual Std. Error	1,050.245  (df = 7854)
F Statistic	$720.960^{***} (df = 5; 7854)$
Note:	*p<0.1; **p<0.05; ***p<0.0

## Model 2:

```
c("Population Density Included?", "Yes", "Yes", "No"),
  c("Average Household Size (v\\_CA21\\_452) Included?", "Yes", "Yes", "No")
),
  title = "Spec 2: Omitted interaction term",
  digits = 3)
```

% Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac at gmail.com % Date and time: Sun, Dec 01, 2024 - 13:27:10

Table 5: Spec 2: Omitted interaction term

	Dependent variable:
	crime_density
(Intercept)	8.053***
	(0.771)
Food Program Density	0.044***
	(0.001)
Population Density	-517.272***
	(22.461)
Average Household Size (v_CA21_452)	252.689***
	(24.682)
Low Income	1,283.891***
	(62.047)
Includes Interaction?	No
Population Density Included?	Yes
Average Household Size (v_CA21_452) Included?	Yes
Observations	7,860
$\mathbb{R}^2$	0.313
Adjusted $R^2$	0.313
Residual Std. Error	1,051.304 (df = 7855)
F Statistic	$895.186^{***} (df = 4; 7855)$
Note:	*p<0.1; **p<0.05; ***p<0

# Model 3:

```
title = "Spec 3: Food Density and Low Income",
digits = 3)
```

% Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac at gmail.com % Date and time: Sun, Dec 01, 2024 - 13:27:10

Table 6: Spec 3: Food Density and Low Income

	Dependent variable:
	crime_density
(Intercept)	9.631***
	(0.897)
Food Program Density	577.872***
	(27.744)
Low Income	383.969***
	(18.802)
Includes Interaction?	No
Population Density Included?	No
Average Household Size (v_CA21_452) Included?	No
Observations	7,860
$\mathbb{R}^2$	0.067
Adjusted $R^2$	0.067
Residual Std. Error	1,225.029 (df = 7857)
F Statistic	$282.621^{***} \text{ (df} = 2; 7857)$
Note:	*p<0.1; **p<0.05; ***p<0.0

% Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac at gmail.com % Date and time: Sun, Dec 01, 2024 - 13:27:10

Table 7: Regression Model Summary (with Persons in Household)

	Dependent variable:
	Crime Density
Food Program Density: Low Income	1.513
	(2.218)
Food Program Density	0.053***
	(0.001)
Population Density	0.225***
	(0.031)
Persons in Household (v_CA21_449)	344.584***
· · · · · · · · · · · · · · · · · · ·	(25.039)
Low Income	9.063***
	(2.374)
Constant	-196.408***
	(25.778)
Observations	7,860
$\mathbb{R}^2$	0.273
Adjusted $R^2$	0.273
Residual Std. Error	1,081.556 (df = 7854)
F Statistic	$590.191^{***} (df = 5; 7854)$
Note:	*p<0.1; **p<0.05; ***p<0.01

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