Intermediate statistics: introduction

Frank Edwards

1/25/2019

School of Criminal Justice, Rutgers - Newark

Basics

Contact: `frank.edwards@rutgers.edu`

Office hours: email for appointments

Course webpage and syllabus:

`https://f-edwards.github.io/intermediate_stats/`

Slack: `https://ru-intermed-stats.slack.com/messages`

Introductions: What are you planning to do with statistical models?

Before we begin

Remember: All models are wrong, some are useful.

What we will cover

- How to explore, visualize, and model diverse kinds of data with an emphasis on generalized linear models
- How to program in R
- Developing a workflow for producing replicable quantitative social science
- Some advanced topics that are relevant for the kinds of data we're dealing with in the course, subject to class interest

Quick assessment of where we’re at with programming

1. Explain what this code does and expected output

```r
k<-2
for(i in 1:10){
  k<-i*k
}
```

2. Explain what this code does and expected output

```
a<-c(1, 2, 3)
b<-c(2, 3, 4)
a*b
```

3. Explain what this code does and expected output

```
whatsitdo<-function(x){
  a<-min(x)
  return(1/a)
}
z<-c(4,5,6)
whatsitdo(z)
```

4. Explain what this code does and expected output

```r
library(dplyr)
dat<-data.frame("var1" = c(1,2,3),
                "var2" = c(4, 5, 6))
dat%>%
  summarise(total = sum(var1 + var2))
```

5. Explain what z is and what m1 is

```
y<-c(1,2,3,4,5)
x<-c(3,4,5,6,7)
z<-solve(t(x)%*%x)%*%t(x)%*%y
m1<-lm(y~x)
```

Discussion on how we could proceed with programming in the course

Self assessment:

Were these problems easy? Hard? Completely foreign? Which parts were most unfamiliar?

Discussion on how we could proceed with programming in the course

Self assessment:

Were these problems easy? Hard? Completely foreign? Which parts were most unfamiliar?

Question for the class:

Would it be helfpul to cover basic programming concepts (i.e. functions, loops)?

Discussion on how we could proceed with programming in the course

Self assessment:

Were these problems easy? Hard? Completely foreign? Which parts were most unfamiliar?

Question for the class:

Would it be helfpul to cover basic programming concepts (i.e. functions, loops)? Using the tidyverse packages?

Discussion on how we could proceed with programming in the course

Self assessment:

Were these problems easy? Hard? Completely foreign? Which parts were most unfamiliar?

Question for the class:

Would it be helfpul to cover basic programming concepts (i.e. functions, loops)? Using the tidyverse packages? Using RMarkdown?

Discussion on how we could proceed with programming in the course

Self assessment:

Were these problems easy? Hard? Completely foreign? Which parts were most unfamiliar?

Question for the class:

Would it be helfpul to cover basic programming concepts (i.e. functions, loops)? Using the tidyverse packages? Using RMarkdown?

Review the syllabus

`https://f-edwards.github.io/intermediate_stats/`

How I will run seminars

- Basic statistical theory

How I will run seminars

- Basic statistical theory
- Applied data analysis and modeling in R

Expectations

- Bring a laptop: we will be writing code in class

Expectations

- Bring a laptop: we will be writing code in class
- Make space for everyone: respect varying levels of comfort with statistics and programming

Expectations

- Bring a laptop: we will be writing code in class
- Make space for everyone: respect varying levels of comfort with statistics and programming
- Come prepared and complete assignments on time

My general approach to data analysis

1. Explore and visualize data

My general approach to data analysis

1. Explore and visualize data
2. Fit models

My general approach to data analysis

1. Explore and visualize data
2. Fit models
3. Assess model fit

My general approach to data analysis

1. Explore and visualize data
2. Fit models
3. Assess model fit
4. Interpret and describe results through simulation

The Generalized Linear Model

The linear model

We know we can model data as:

$$y = \beta_0 + \beta_1 x_1 \cdots \beta_n x_n + \varepsilon$$

The linear model

We know we can model data as:

$$y = \beta_0 + \beta_1 x_1 \cdots \beta_n x_n + \varepsilon$$

Or, more succinctly:

$$y = \mathbf{X}\beta + \varepsilon$$

The linear model

We know we can model data as:

$$y = \beta_0 + \beta_1 x_1 \cdots \beta_n x_n + \varepsilon$$

Or, more succinctly:

$$y = \mathbf{X}\beta + \varepsilon$$

Where the likelihood for the outcome conditional on the data takes the form:

$$Y|X \sim Normal(\mu, \sigma^2)$$

Generalzing the linear model

The linear model:

$$Y|X \sim Normal(\mu, \sigma^2)$$

Can be written as a more general formulation for a likelihood function f

$$Y|X \sim f(\mu, \sigma^2)$$

Now we can extend the (very) useful linear model to data with discrete outcomes

Generalizing the linear model

A linear predictor η:

$$\eta = \mathsf{X}\beta$$

A link function g

$$g(E(Y|X)) = \eta$$

A mean expectation $E(Y|X) = \mu$

$$\mu = g^{-1}(\eta)$$

From OLS to GLM

OLS:

$$Y|X \sim Normal(\mu, \sigma^2)$$

GLM:

$$Y|X \sim f(\mu, \sigma^2)$$

Diverse likelihood functions

- Binary data: linear probability and logistic models

Diverse likelihood functions

- Binary data: linear probability and logistic models
- Categorical data: Multinomial model

Diverse likelihood functions

- Binary data: linear probability and logistic models
- Categorical data: Multinomial model
- Count data: Poisson and negative binomial models

Diverse likelihood functions

- Binary data: linear probability and logistic models
- Categorical data: Multinomial model
- Count data: Poisson and negative binomial models
- Positive continuous data: Gamma model

Getting started: software

Required installations

All software we are using is free and open source.

Install R:

`https://cran.r-project.org/`

Install RStudio:

`https://www.rstudio.com/products/rstudio/download/`

Recommended software: Git and GitHub

Git and GitHub are powerful tools for backing up and sharing your research.

All course materials, source code, and most of my research are hosted on GitHub (`https://github.com/f-edwards`).

Install Git:

`https://git-scm.com/`

Set up a GitHub account:

`https://github.com/`

Using GitHub for social science:

`https://happygitwithr.com/`

Recommended software: LaTeX

LaTeX is a powerful typesetting tool that works well with RMarkdown. It makes very attractive academic papers and slides.

Install it here: *Install TexLive*:

`https://tug.org/texlive/acquire-netinstall.html`

Questions so far?

Break

Returning to the linear model

What do we know about the linear regression model?

$$y = \mathbf{X}\beta + \varepsilon$$

$$\varepsilon \sim Normal(0, \sigma^2)$$

Review

1. What forms can y take?
2. What assumptions does the linear regression model require?
3. What are some contexts where the linear regression model can be misleading?

Let’s build some models to review

Two ways to access course data

- All data is accessible through the the course website (see the data link, or data folder on the GitHub page)
- *Recommended approach:* In a terminal (terminal.app on mac, Git Bash on windows):

\texttt{git clone
`https://github.com/f-edwards/intermediate_stats.git`}

Before beginning your work each session, pull updates I've pushed to the repo with:

```
git pull
```

Now you have an intermediate_stats folder with all code, slides, and data. Data is in intermediate_stats/data

Read in

```
#library(tidyverse)
### directly from the web
cj_budgets<-read_csv("https://github.com/f-edwards/intermediate_
### from a project directory root
#setwd("C:/intermediate_stats") # set working directory
#cj_budgets<-read_csv("./data/revenue_dat.csv")
```

About the data

Data are for an ongoing research project I'm working on. It's real, and can be a bit messy!

It documents police involved deaths, demographics, and local government budgets at the county-level for two time periods, 2007-11 and 2012-16. Datasets used include Fatal Encounters, American Community Survey, Annual Survey of State and Local Government Finance, and Uniform Crime Reports.

Full code for the project is up at:

`https://github.com/f-edwards/police-mort-revenue`

`merge.r` contains the code to make this merged file from a variety of source files (available if you want the raw data).

Evaluate the structure of the data

```
head(cj_budgets)
```

```
## # A tibble: 6 x 33
##   year_range fips_st fips_cnty deaths exp_tot exp_correction
##   <chr>      <chr>   <chr>      <dbl>   <dbl>          <dbl>
## 1 2007-2011  01      001            3  4.97e7        2101800
## 2 2007-2011  01      005            1  2.86e7        1037880.
## 3 2007-2011  01      007            0  1.30e7          80600
## 4 2007-2011  01      009            0  3.66e7        1703760
## 5 2007-2011  01      011            0  1.09e7              0
## 6 2007-2011  01      013            1  3.05e7         487320
## # ... with 26 more variables: exp_welfare <dbl>, rev_tot <dbl
## #   rev_fines <dbl>, rev_gen_ownsource <dbl>, rev_int_gov <db
## #   rev_prop_tax <dbl>, rev_tax <dbl>, pop_tot <dbl>,
## #   pop_pct_men_15_34 <dbl>, pop_wht <dbl>, pop_blk <dbl>, po
## #   pop_api <dbl>, pop_lat <dbl>, pop_pct_pov <dbl>,
## #   pop_pct_deep_pov <dbl>, pop_med_income <dbl>, pop_pc_inco
```

Evaluate the structure of the data

```
nrow(cj_budgets)
```

```
## [1] 4286
```

```
table(cj_budgets$year_range)
```

```
##
## 2007-2011 2012-2016
##      2308      1978
```

Evaluate the structure of the data

```
names(cj_budgets)
```

```
##  [1] "year_range"         "fips_st"            "fips_cnty"
##  [4] "deaths"             "exp_tot"            "exp_correction"
##  [7] "exp_police"         "exp_welfare"        "rev_tot"
## [10] "rev_fines"          "rev_gen_ownsource" "rev_int_gov"
## [13] "rev_prop_tax"       "rev_tax"            "pop_tot"
## [16] "pop_pct_men_15_34" "pop_wht"            "pop_blk"
## [19] "pop_ami"            "pop_api"            "pop_lat"
## [22] "pop_pct_pov"        "pop_pct_deep_pov"  "pop_med_income"
## [25] "pop_pc_income"      "violent.yr"         "property.yr"
## [28] "murder.yr"          "ft_sworn"           "cbsa"
## [31] "metroname"          "dissim_bw"          "dissim_wl"
```

Descriptives

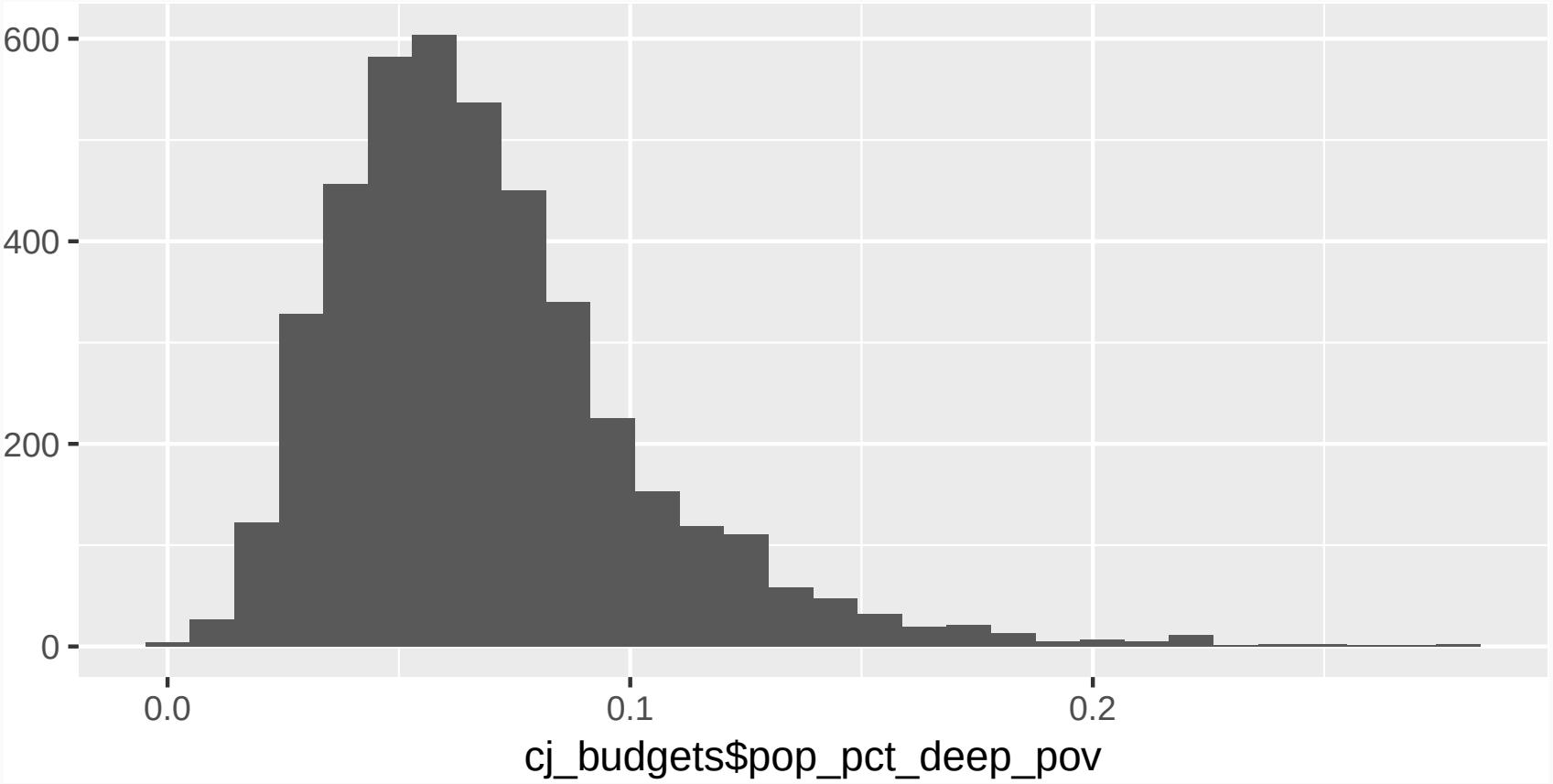
```
summary(cj_budgets$pop_pct_deep_pov)
```

```
##     Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.00000 0.04553 0.06285 0.06884 0.08442 0.27901
```

Visualize the distribution of a variable

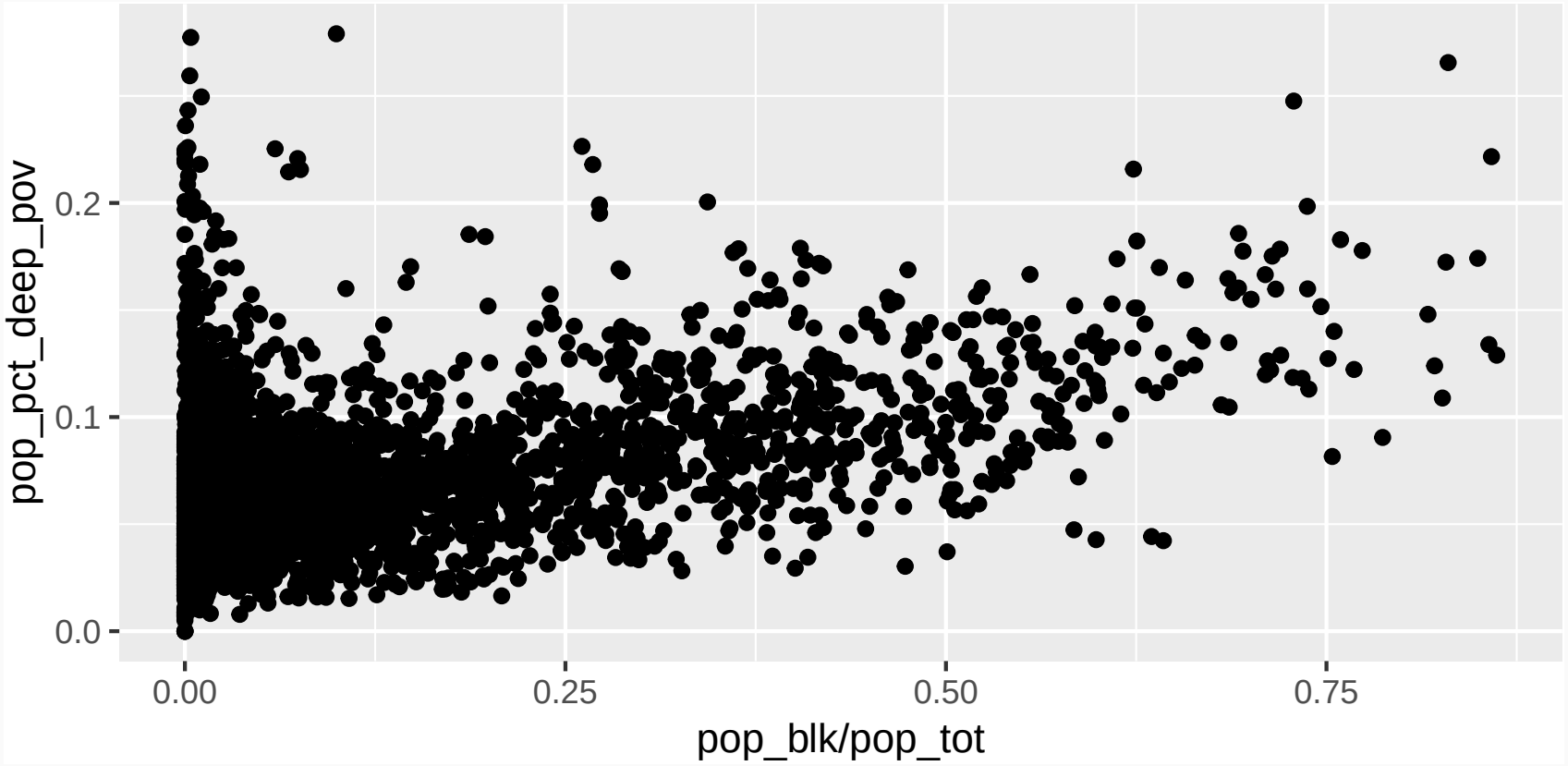
Call individual variables (columns) in a data frame with $, like \texttt{USArrests$Murder}

```
qplot(cj_budgets$pop_pct_deep_pov)
```



Visualize a bivariate relationship

```
qplot(pop_blk/pop_tot,
      pop_pct_deep_pov,
      data = cj_budgets)
```



Fitting a linear model

```
model_1<-lm(pop_pct_deep_pov ~
              I(pop_blk/pop_tot),
            data =cj_budgets)
```

Display the model fit

```
summary(model_1)
```

```
##
## Call:
## lm(formula = pop_pct_deep_pov ~ I(pop_blk/pop_tot), data = cj
##
## Residuals:
##       Min         1Q     Median         3Q        Max
## -0.079709 -0.019343 -0.004579  0.013753  0.217773
##
## Coefficients:
##                     Estimate Std. Error t value Pr(>|t|)
## (Intercept)        0.0591712  0.0005603  105.61   <2e-16 ***
## I(pop_blk/pop_tot) 0.0977188  0.0030884   31.64   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '
##
```

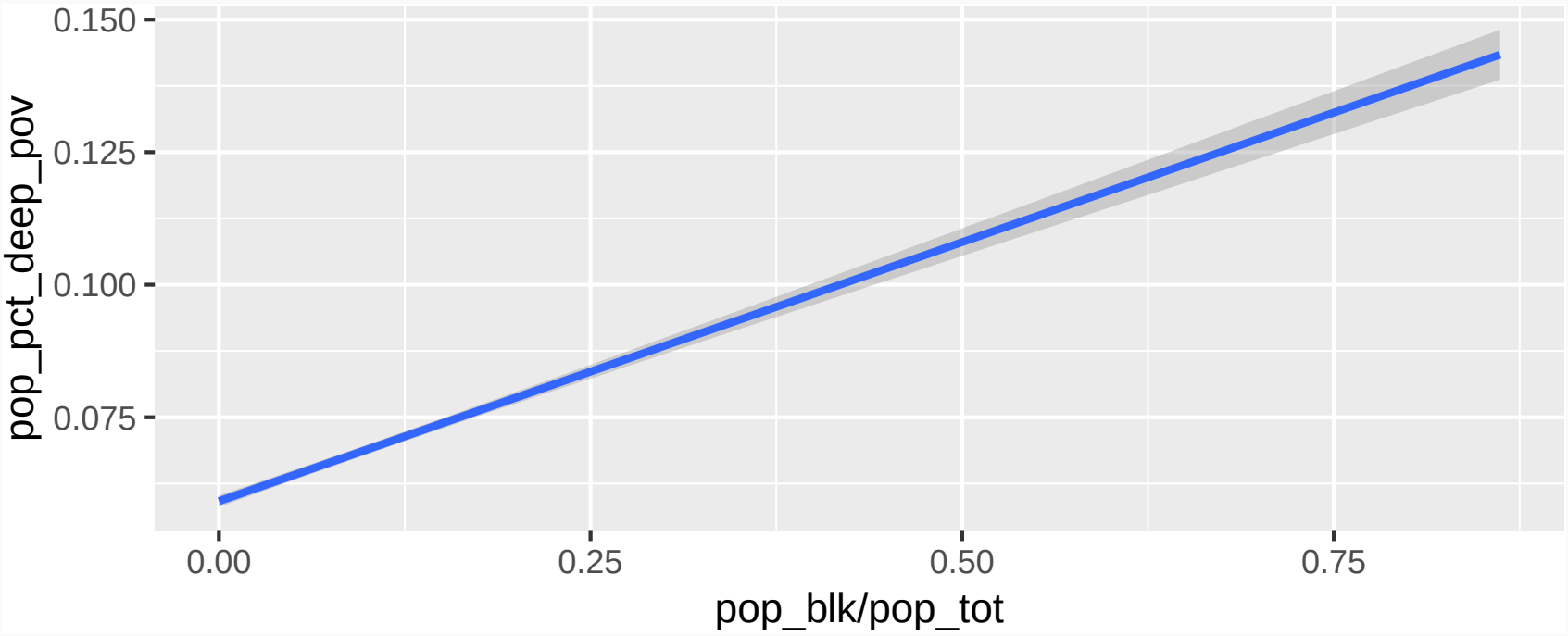
Display the model fit (nicer)

```
library(broom)
tidy(model_1)
```

```
## # A tibble: 2 x 5
##   term                estimate std.error statistic   p.value
##   <chr>                  <dbl>     <dbl>     <dbl>     <dbl>
## 1 (Intercept)           0.0592  0.000560      106.  0.
## 2 I(pop_blk/pop_tot)    0.0977  0.00309        31.6 1.22e-197
```

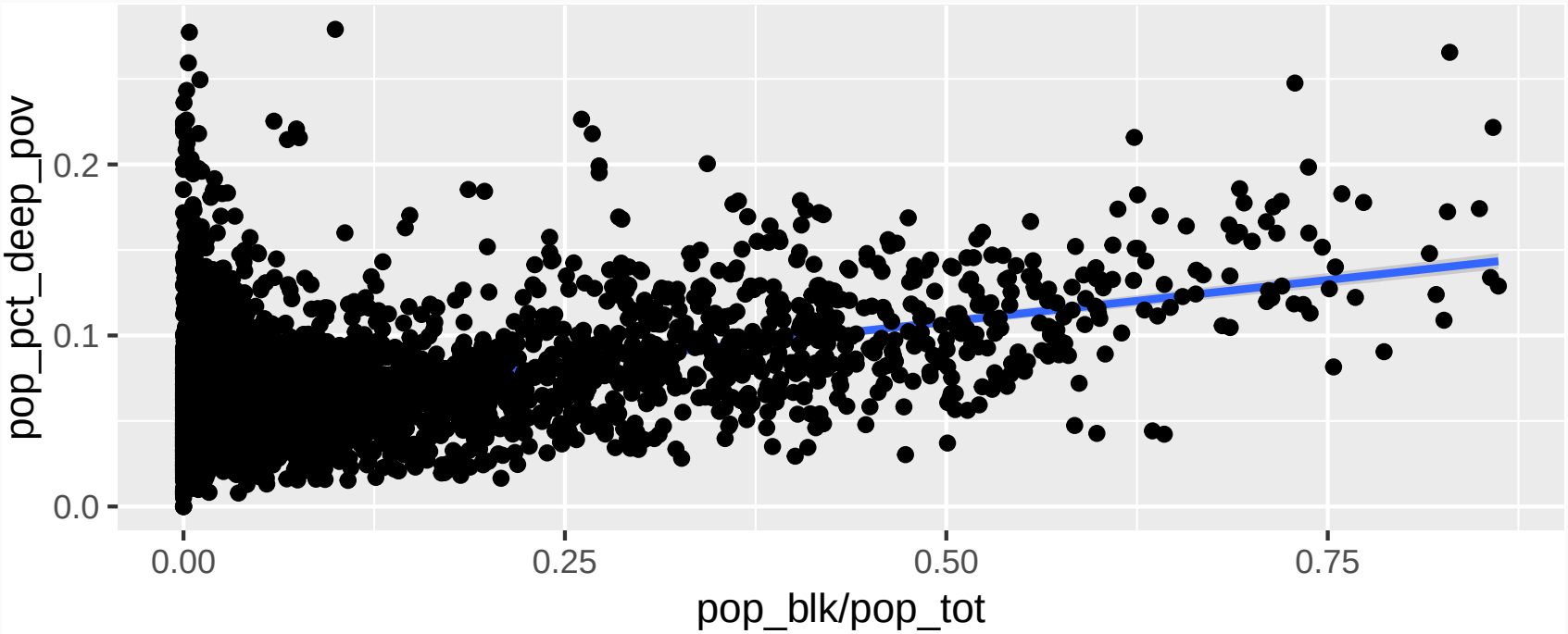
Visualize the model fit

```
library(ggplot2)
ggplot(cj_budgets,
       aes(x=pop_blk/pop_tot, y=pop_pct_deep_pov))+
  geom_smooth(method = "lm",
              formula = y~x)
```



Visualize the model fit (against the data)

```
library(ggplot2)
ggplot(cj_budgets,
       aes(x=pop_blk/pop_tot, y=pop_pct_deep_pov))+
  geom_smooth(method = "lm",
              formula = y~x) +
  geom_point()
```



Can we fit a better model?