Using R for Analytic Graphs: Learn How Data Visualization Can Improve Interpretation in Social Work Research

Saturday, November 01, 2014

Why Use R?

- Free
- Open Source
- ► Easy Collaboration
- ► Replicable Research

Why Wouldn't You Use R?

Steep(er) learning curve compared to, say, Excel or SPSS. This matters a lot if

- You run statistics rarely.
- You want a point and click interface.

Where Can you Get R?

- ► CRAN
- Our Thumb Drives

Where Are We Going Today?

- Graphing Model Results
- Graphing Other Things

Graphing Model Results (Basic Algorithm)

- 1. Choose a counterfactual x_c .
- 2. Estimate a model to get a vector of parameters $\hat{\beta}$ and the associated variance-covariance matrix, $\hat{\mathbf{V}}$.
- 3. Draw several $\tilde{\boldsymbol{\beta}}$ from $\mathcal{N}(\hat{\boldsymbol{\beta}}, \hat{\boldsymbol{V}})$, where \mathcal{N} is a mulivariate normal distribution.
- 4. Calculate expected outcomes based on model parameters for all of your draws from \mathcal{N} .
- 5. Calculate summary statistics for each level of x_c .

This approach will work for most of the models that social welfare researchers tend to encounter.

A Practical Example - Background

Research Question

How does a child's probability of exiting the foster care system vary by child characteristics?

Multiple Permanency Outcomes

Requires that we estimate a mulinomial logistic regression model.

Data in Question

- ▶ 500 children entering out-of-home care in late 2007.
- Children's parent's were surveyed once in 2007. The survey results were then linked to administrative data which faciliated a longitudinal follow-up.
- ▶ Data have been jittered and randomly sampled from a larger set of data to mask the identity of subjects. The data used here do not reflect the data of individual subjects.



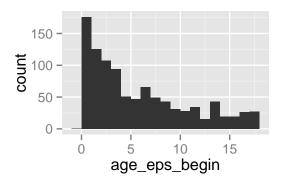
A practical example - Choose a counterfactual x_c .

Load the data

```
dat <- read.csv("dat.csv")</pre>
```

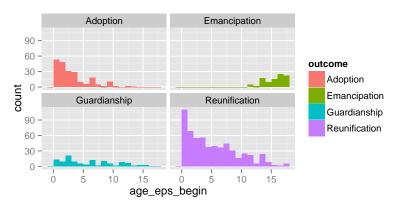
A practical example - Choose a counterfactual x_c .

```
#looking at age of child at episode begin
require(ggplot2)
ggplot(dat, aes(x=age_eps_begin)) +
  geom_histogram(binwidth = 1)
```



A practical example - Choose a counterfactual x_c .

```
#looking at age of child at episode begin by outcome
ggplot(dat, aes(x=age_eps_begin, fill=outcome)) +
  geom_histogram(binwidth = 1) +
  facet_wrap(~ outcome)
```



Need to estimate a statistical model to get

- 1. A vector of parameters $\hat{\boldsymbol{\beta}}$, and
- 2. The associated variance-covariance matrix, $\hat{\boldsymbol{V}}$.

Prep the data

```
# easy to load external packages
# install.packages("nnet") # install once
require(nnet)
                           # load every time
# relevel our outcome variable
dat$outcome_rl <- relevel(dat$outcome</pre>
                            , ref = "Emancipation")
# recode to numeric
dat$outcome_rl <- as.numeric(dat$outcome rl)</pre>
```

Run the model

```
## # weights: 16 (9 variable)
## initial value 1386.294361
## iter 10 value 931.103300
## iter 20 value 860.375750
## final value 860.374425
## converged
```

Display of summary the model

```
model
```

```
## Call:
## multinom(formula = outcome_rl ~ age_eps_begin + eps_ranl
##
      Hess = TRUE
##
## Coefficients:
    (Intercept) age_eps_begin eps_rank
##
## 2
    11.457 -1.0281 -0.10995
        9.798
## 3
                 -0.8393 0.05195
## 4
    11.597 -0.8691 0.07150
##
## Residual Deviance: 1721
## ATC: 1739
```

[1] 11.5972 -0.8691 0.0715

Extract a vector of parameters $\hat{\boldsymbol{\beta}}$

```
#run the multinomial model
pe <- model$wts[c(6,7,8,10,11,12,14,15,16)]
pe[1:3]
## [1] 11.457 -1.028 -0.110
pe[4:6]
## [1] 9.79767 -0.83931 0.05195
pe[7:9]
```

Extract the associated variance-covariance matrix, $\hat{\boldsymbol{V}}$

#run the multinomial model vc <- solve(model\$Hess)</pre>

A practical example - Draw several $\tilde{\boldsymbol{\beta}}$ from $\mathcal{N}(\hat{\boldsymbol{\beta}},\ \hat{\boldsymbol{V}})$.

```
#load a package which contains a multivariate normal
#sampling function
require(MASS)
#assign a variable for the number of simulations
sims <- 10000
#draw the indicates number of beta simulates
#using our extracted model data
simbetas <- mvrnorm(sims,pe,vc)</pre>
```

A practical example - Last two steps. . .

- ightharpoonup Calculate expected values for all of your draws from $\mathcal N$, and
- ▶ Calculate summary statistics for each level of x_c .
- ► Specific calculations are beyond the scope of this presentation
- ▶ But the simcf package from Chris Adolph (political scientist at the University of Washington) will do them for us!

A practical example - Last two steps

Get data read for simcf

▶ Re-arrange simulates to array format

```
simb <- array(NA, dim = c(sims,3,3))
simb[,,1] <- simbetas[,1:3]
simb[,,2] <- simbetas[,4:6]
simb[,,3] <- simbetas[,7:9]</pre>
```

Specify range of counterfactual values

```
agerange <- seq(0,17,by=0.1)
```

A practical example - Last two steps

Get data read for simcf

► Load simcf and use the cfFactorial() function to set specific values for simulation.

▶ Run the simulation (this is where the last two steps are really performed).

```
test_sims <- mlogitsimev(xhyp,simb,ci=0.95)</pre>
```

Get the data ready to graph

```
y <- as.vector(test_sims$pe[,1:4])
x <- rep(1:length(agerange), 4)
lower <- as.vector(test_sims$lower[,1:4,])</pre>
upper <- as.vector(test sims$upper[,1:4,])
Outcome <- c(rep("Adoption", length(agerange))
                  ,rep("Guardianship"
                       ,length(agerange))
                  ,rep("Reunification"
                       ,length(agerange))
                  ,rep("Emancipation"
                       ,length(agerange)))
```

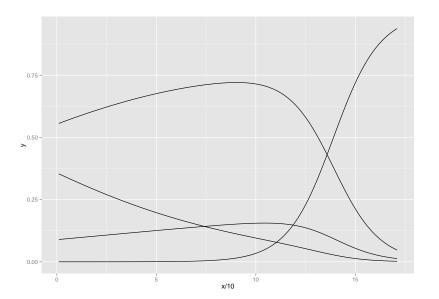
Get the data ready to graph

```
dat_sim_plot <- data.frame(y,x,lower,upper,Outcome)</pre>
```

Graph the data!

```
p1 <- ggplot(dat_sim_plot
    ,aes(x=x/10, y=y, group=Outcome)) +
        geom_line()</pre>
```

Graph the data!



Make it Pretty!

Make it Pretty

