

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

Saturday, January 18, 2015

# 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 model parameters  $\hat{\beta}$  and the variance-covariance matrix,  $\hat{\mathbf{V}}$ .
3. Draw several  $\tilde{\beta}$  from  $\mathcal{N}(\hat{\beta}, \hat{\mathbf{V}})$ , where  $\mathcal{N}$  is a multivariate normal distribution.
4. Calculate expected outcomes based on model parameters for all draws from  $\mathcal{N}$ .
5. Calculate summary statistics for each level of  $x_c$ .

This approach will work for most models that social welfare researchers use.

# A Practical Example - Background

- ▶ Research Question: How does a child's probability of exiting foster care vary by child characteristics?
- ▶ Multiple Permanency Outcomes: Requires that we estimate a multinomial logistic regression model.
- ▶ Data: Children ( $n = 619$ ) entering out-of-home care in late 2007. Children's parents' surveyed in 2007 and linked to administrative data to facilitate follow-up.
- ▶ NOTE: These data are simulated because to protect client confidentiality.

# A practical example - Choose a counterfactual $x_c$ .

## Getting Data Into R

- ▶ R can import and export most formats: `read.dta()`, `read.spss()`, `read.csv()`, `read.dbf()`
- ▶ Use the `foriegn()` library to import SPSS, Stata, and DBase formats.

```
dat <- read.csv("dat.csv")
```

-To export use the `write.csv()` command. # 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)
```



# A practical example - Estimate a model.

Need to estimate a statistical model to get

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

# A practical example - Estimate a model.

## Prep the data

```
# easy to load external packages  
# install.packages("nnet") # install once  
require(nnet) # load every time  
  
# recode data  
levels(dat$outcome) <- c("Adoption", "Emancipation"  
                        , "Guardianship", "Reunification")  
  
# relevel our outcome variable  
dat$outcome_rl <- relevel(dat$outcome  
                        , ref = "Emancipation")  
  
# recode to numeric  
dat$outcome_rl <- as.numeric(dat$outcome_rl)
```

# A practical example - Estimate a model.

## Run the model

```
# run the multinomial model  
model <- multinom(outcome_rl ~ age_eps_begin +  
                  eps_rank  
                  ,data = dat  
                  ,Hess = TRUE)
```

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

# A practical example - Estimate a model.

## Display of summary the model

```
model
```

```
## Call:
## multinom(formula = outcome_rl ~ age_eps_begin + eps_rank,
##          Hess = TRUE)
##
## Coefficients:
##      (Intercept) age_eps_begin      eps_rank
## 2      11.457365      -1.0280750 -0.10995325
## 3       9.797665      -0.8393067  0.05195097
## 4      11.597181      -0.8691345  0.07149574
##
## Residual Deviance: 1720.749
## AIC: 1738.749
```

## A practical example - Estimate a model.

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

```
#run the multinomial model
```

```
pe <- model$wts[c(6,7,8,10,11,12,14,15,16)]  
pe[1:3]
```

```
## [1] 11.4573653 -1.0280750 -0.1099532
```

```
pe[4:6]
```

```
## [1] 9.79766546 -0.83930667 0.05195097
```

```
pe[7:9]
```

```
## [1] 11.59718150 -0.86913446 0.07149574
```

## A practical example - Estimate a model.

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

```
#run the multinomial model  
vc <- solve(model$Hess)
```

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

```
#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)
```



## A practical example - Last two steps. . .

- ▶ 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]
```

- ▶ 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.

```
require(simcf)
xhyp <- cfFactorial(age = agerange
                    ,ep_rank = mean(dat$eps_rank))
```

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

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

## 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,])  
  
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)))
```

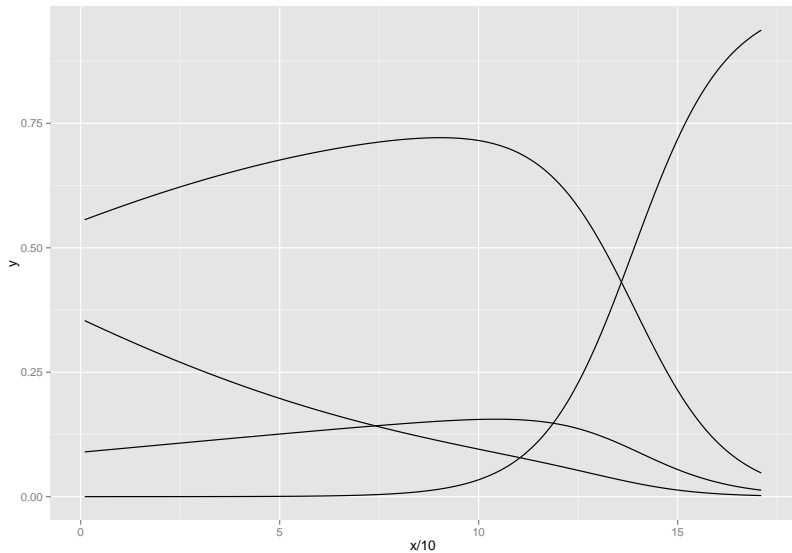
## Get the data ready to graph

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

# Graph the data!

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

# Graph the data!



# Make it Pretty!

```
p2 <- ggplot(dat_sim_plot
  ,aes(x=x/10, y=y, group=Outcome)) +
  geom_line(size=1, alpha=.5) +
  geom_ribbon(aes(ymin=lower
                  ,ymax=upper
                  ,fill=Outcome), alpha=.5) +
  ylab("Pr(Outcome|Age,Prior Episodes)") +
  xlab("Age at Entry into Foster Care") +
  theme_bw()
```



# Make it Pretty

