

# Answers for Plotting

*Hao Wang*

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## 1. Load essential packages

```
library(VGAM)
library(car)
library(MASS)
library(effects)
library(ggplot2)
library(Zelig)
library(ZeligChoice)
library(readstata13)
play<- read.dta13('https://github.com/haowang666/POS603-Lab/blob/master/Lab%206/stdSingh.dta?raw=true')
```

## 2. Run Zelig estimates

```
zlogit <- zelig(voted ~severity*age + polinfrel + income + efficacy + partyID +
               dist_magnitude + enep + vicmarg_dist +
               parliamentary + development, model = "logit", data=play, cite = FALSE)
summary(zlogit)
```

```
## Model:
##
## Call:
## z5$zelig(formula = voted ~ severity * age + polinfrel + income +
##          efficacy + partyID + dist_magnitude + enep + vicmarg_dist +
##          parliamentary + development, data = play)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5291   0.2564   0.4129   0.5869   1.7531
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    2.105223   0.012538 167.911 < 2e-16
## severity        0.637247   0.020068  31.754 < 2e-16
## age             0.303940   0.011051  27.504 < 2e-16
## polinfrel       0.333097   0.010654  31.264 < 2e-16
## income          0.220452   0.010350  21.300 < 2e-16
## efficacy        0.362777   0.009043  40.116 < 2e-16
## partyID         0.424113   0.010723  39.553 < 2e-16
## dist_magnitude  0.185052   0.012521  14.779 < 2e-16
## enep            -0.107988   0.014018  -7.703 1.32e-14
## vicmarg_dist    -0.035428   0.010394  -3.408 0.000653
## parliamentary   0.137085   0.011165  12.278 < 2e-16
## development     0.311358   0.012371  25.168 < 2e-16
```

```
## severity:age    -0.119769    0.016218   -7.385 1.52e-13
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 77479   on 92740   degrees of freedom
## Residual deviance: 67443   on 92728   degrees of freedom
## AIC: 67469
##
## Number of Fisher Scoring iterations: 6
##
## Next step: Use 'setx' method
```

## Check Ranges

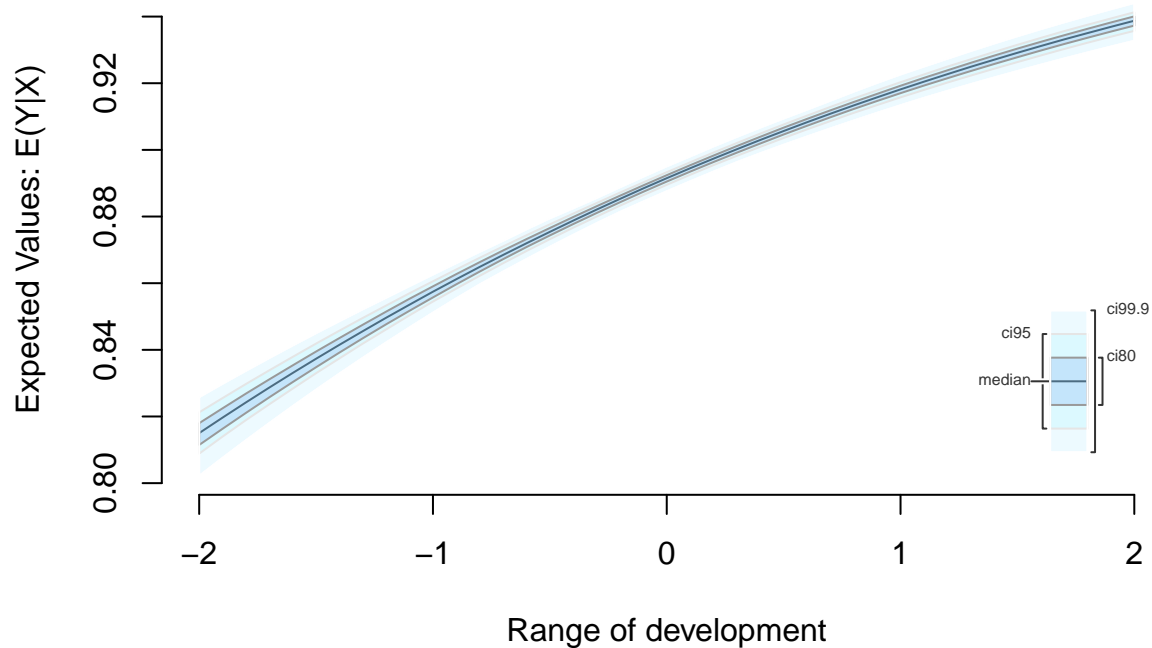
```
#check range
summary(play$development)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -1.8040 -0.7614   0.2206   0.0000   0.7541   1.9530
```

## Create data and run simulation

```
#41 Xs
x.sim <- setx(zlogit, development = seq(from = -2, to = 2, by = 0.1))
s.out <- sim(zlogit, x = x.sim)
```

```
#plot(s.out)
ci.plot(s.out, qi = "ev")
```



## If you are unhappy with the default plot...

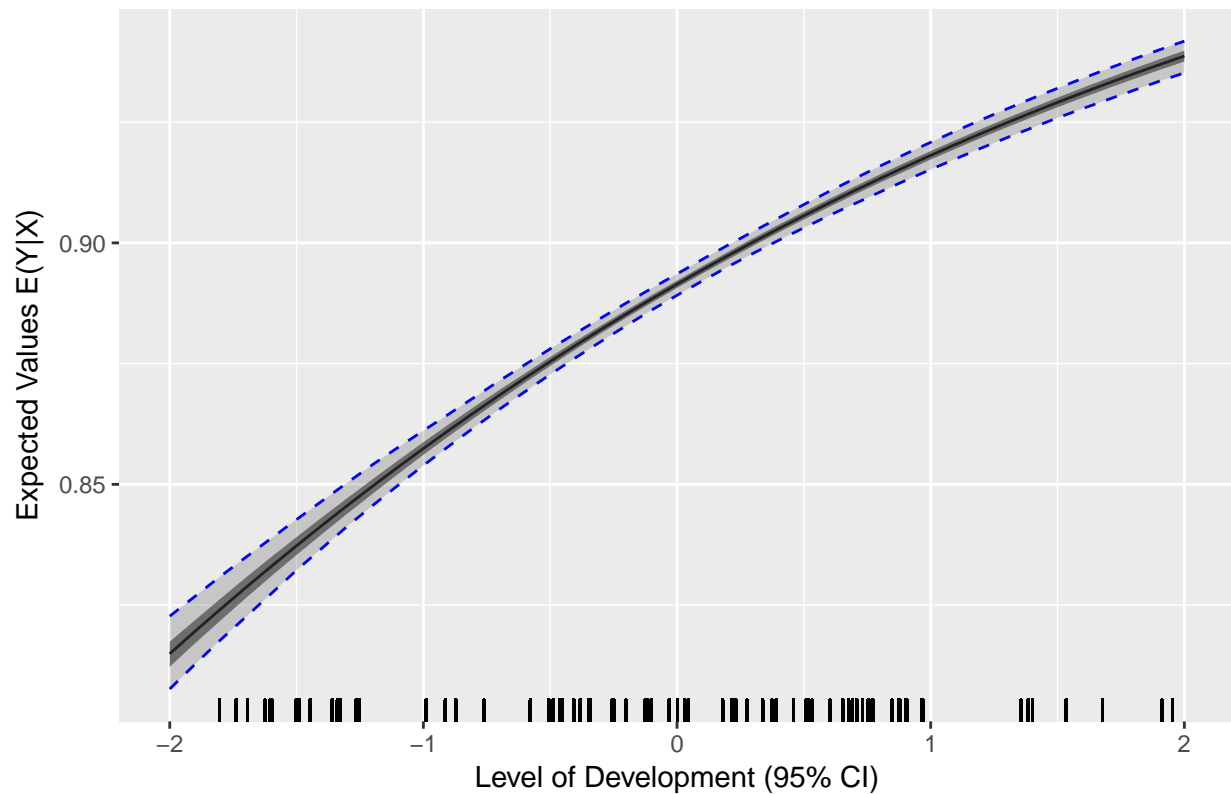
We need to extract info from zelig simulation

```
## Ranges of values
development = seq(from = -2, to = 2, by = 0.1)

# Extract expected values from simulations
# ev is expected values
phat<- s.out$get_qi(qi='ev', xvalue = 'range')
# collapse list to dataframe
phat <- as.data.frame(matrix(unlist(phat), nrow =1000))
# apply function to column
a<- apply(phat, 2, quantile, probs = c(0.025, 0.975, 0.25, 0.75))
low <- a[1,]
high <- a[2,]
qt.1 <- a[3,]
qt.3 <- a[4,]
fit <- apply(phat, 2, mean)
plotdata <- as.data.frame(cbind(low, high, fit, qt.1, qt.3))
plotdata$development <- development

#in ggplot2
ggplot() +
  geom_line(data = plotdata, aes(x = development, y =fit)) +
  geom_line(data = plotdata, aes(x = development, y =high), linetype="dashed", color="blue") +
  geom_line(data = plotdata, aes(x = development, y =low), linetype="dashed", color="blue") +
  geom_rug(data = play, aes(x= development)) +
  xlab("Level of Development (95% CI)") +
  ylab("Expected Values E(Y|X)") +
  ggtitle("Marginal Effect of Development")+
  geom_ribbon(data=plotdata, aes(x=development, ymin=low, ymax=high), alpha=0.2) +
  geom_ribbon(data=plotdata, aes(x=development, ymin=qt.1, ymax=qt.3), alpha=0.6)
```

## Marginal Effect of Development



### 3. Or use vanilla GLM code

```
glm <- glm(formula = voted ~ severity*age + polinfrel + income
           + efficacy + partyID +
           dist_magnitude + enep + vicmarg_dist +
           parliamentary + development,
           data = play, family = binomial(link = "logit"))
development <- seq(-2, 2, by = .1)
attach(play)
newdata <- as.data.frame(cbind(1, development, mean(severity),
                              mean(age), mean(polinfrel), mean(income),
                              mean(efficacy), mean(partyID),
                              mean(vicmarg_dist), mean(parliamentary),
                              mean(dist_magnitude), mean(enep)))

colnames(newdata) <- c("constant", "development",
                      "severity", "age", "polinfrel", "income",
                      "efficacy", "partyID", "vicmarg_dist",
                      "parliamentary", "dist_magnitude", "enep")

#response offers expected value
pr <- predict(glm, newdata = newdata, se.fit = TRUE, type = "response")
#what if you missed the command type="response"?
#you get linear prediction results
pr2 <- predict(glm, newdata = newdata)
```

```
fit2 <- 1/(1+exp(- pr2))
cor(pr$fit, fit2)
```

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

```
detach(play)
```

## Plot it

```
plotdata <- as.data.frame(pr$fit)
colnames(plotdata)<-c("fit")
plotdata$se <- pr$se.fit
plotdata$low <- plotdata$fit - 1.96*plotdata$se
plotdata$high <- plotdata$fit + 1.96*plotdata$se
plotdata$development <- development

ggplot() +
  geom_line(data = plotdata, aes(x = development, y =fit)) +
  geom_line(data = plotdata, aes(x = development, y =high), linetype="dashed", color="blue") +
  geom_line(data = plotdata, aes(x = development, y =low), linetype="dashed", color="blue") +
  geom_rug(data = play, aes(x= development)) +
  xlab("Level of Development (95% CI)") +
  ylab("Expected Values E(Y|X)") +
  ggtitle("Marginal Effect of Development")+
  geom_ribbon(data=plotdata, aes(x=development, ymin=low, ymax=high), alpha=0.5)
```

