

Quantitative Methods III

Homework 1

Alexander Herzog

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(I) Huber et al. (2005) data

Note: I named the Huber et al. data set `data.R` which is a really bad name as `.R` is usually reserved for syntax files. The convention is to use the extension `.Rdata`

Load Gelman's "arm" package.

```
> library(arm)
```

Change working directory

```
> setwd("~/NYU/QuantIII/hmwk01/")
```

a) Get the data for the huber et al study into R

```
> load("huber_et_al_2005.Rdata")
```

b) Run some individual country logits

b.1) Individual country regressions for Germany, Ireland, and Slovenia

```
> pid.germany <- glm(simplepid ~ education + age1 + age2 + male,
+                   subset = country == "Germany",
+                   family=binomial(link="logit"))

> display(pid.germany)
glm(formula = simplepid ~ education + age1 + age2 + male, family = binomial(link = "logit"),
    subset = country == "Germany")
      coef.est coef.se
(Intercept) -0.76   0.08
education    0.23   0.03
age1         0.25   0.05
age2        -0.02   0.02
male         0.40   0.10
---
n = 1906, k = 5
residual deviance = 2362.6, null deviance = 2481.3 (difference = 118.7)

> pid.ireland <- glm(simplepid ~ education + age1 + age2 + male,
```

```

+             subset = country == "Ireland",
+             family=binomial(link="logit"))

> display(pid.ireland)
glm(formula = simplepid ~ education + age1 + age2 + male, family = binomial(link = "logit"),
     subset = country == "Ireland")
      coef.est coef.se
(Intercept) -1.22   0.08
education    0.03   0.03
age1         0.25   0.04
age2        -0.01   0.01
male         0.28   0.09
---
n = 2291, k = 5
residual deviance = 2658.5, null deviance = 2727.3 (difference = 68.8)

> pid.slovenia <- glm(simplepid ~ education + age1 + age2 + male,
+                    subset = country == "Slovenia",
+                    family=binomial(link="logit"))

> display(pid.slovenia)
glm(formula = simplepid ~ education + age1 + age2 + male, family = binomial(link = "logit"),
     subset = country == "Slovenia")
      coef.est coef.se
(Intercept) -1.62   0.10
education    0.17   0.03
age1         0.22   0.05
age2        -0.01   0.02
male         0.46   0.12
---
n = 1869, k = 5
residual deviance = 1833.9, null deviance = 1898.4 (difference = 64.6)

```

b.2) All countries with regression of country b's for education on a single group level predictor (= "secret weapon")

```

> j <- 1
> coefs <- NA

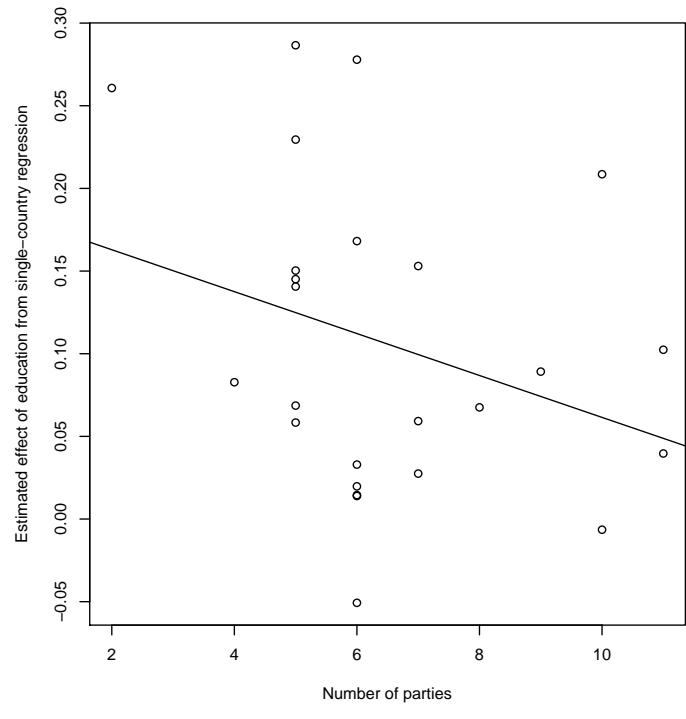
> for(i in unique(country)) {
+   reg.i <- glm(simplepid ~ education + age1 + age2 + male,
+               subset = country == i,
+               family=binomial(link="logit"))
+   coefs[j] <- reg.i$coef[2]
+   j <- j + 1
+ }

> no.parties.aggregated <- tapply(no.parties.gt.1, country, mean)

> reg.coef <- lm(coefs ~ no.parties.aggregated)

> plot(no.parties.aggregated,coefs,
+      xlab="Number of parties", ylab="Estimated effect of education
+      from single-country regression")
> abline(reg.coef)

```



c) Take the full data set and run with interactions

I specify three different models: a complete-pooling model, a no-pooling model, and a no-pooling model with interactions between country dummies and education.

c.1) Complete-pooling model

```
> pooling <- glm(simplepid ~ education + age1 + age2 + male , family=binomial(link="logit"))

> display(pooling)
glm(formula = simplepid ~ education + age1 + age2 + male, family = binomial(link = "logit"))
      coef.est coef.se
(Intercept) -0.23    0.01
education    0.12    0.01
age1         0.19    0.01
age2        -0.02    0.00
male         0.16    0.02
---
n = 56073, k = 5
residual deviance = 76210.1, null deviance = 77629.6 (difference = 1419.6)
```

c.2) No-pooling model: Varying-intercept model without interactions

```
> no.pooling <- glm(simplepid ~ education + age1 + age2 + male +
+                   as.factor(country),
+                   family=binomial(link="logit"))

> display(no.pooling)
glm(formula = simplepid ~ education + age1 + age2 + male + as.factor(country),
     family = binomial(link = "logit"))
      coef.est coef.se
(Intercept)      1.45  0.07
education         0.10  0.01
age1              0.20  0.01
age2            -0.02  0.00
male              0.17  0.02
as.factor(country)Belgium      1.31  0.12
as.factor(country)Bulgaria    -1.92  0.09
as.factor(country)Canada     -1.65  0.08
as.factor(country)Czech Rep  -1.51  0.08
as.factor(country)Denmark    -1.68  0.08
as.factor(country)France     -1.69  0.09
as.factor(country)Germany    -2.15  0.08
as.factor(country)Hungary    -1.90  0.08
as.factor(country)Iceland    -1.57  0.08
as.factor(country)Ireland    -2.57  0.08
as.factor(country)Israel     -1.12  0.08
as.factor(country)Japan      -2.16  0.09
as.factor(country)Lithuania  -2.29  0.10
as.factor(country)Netherlands -2.58  0.08
as.factor(country)New Zealand -1.39  0.07
as.factor(country)Norway     -1.82  0.07
as.factor(country)Poland     -1.65  0.08
as.factor(country)Portugal   -1.56  0.08
as.factor(country)Slovenia   -2.91  0.09
as.factor(country)Spain     -1.80  0.08
```

```

as.factor(country)Sweden      -1.62    0.08
as.factor(country)Switzerland -2.06    0.08
as.factor(country)UK          -1.74    0.08
as.factor(country)USA         -1.46    0.09
---
n = 56073, k = 29
residual deviance = 71308.5, null deviance = 77629.6 (difference = 6321.1)

```

c.3) No-pooling model with interactions for 'education'

```

> no.pooling.int <- glm(simplepid ~ education + age1 + age2 + male +
+                       as.factor(country) + as.factor(country):education,
+                       family=binomial(link="logit"))

> display(no.pooling.int)
glm(formula = simplepid ~ education + age1 + age2 + male + as.factor(country) +
     as.factor(country):education, family = binomial(link = "logit"))

```

	coef.est	coef.se
(Intercept)	1.49	0.07
education	-0.04	0.04
age1	0.20	0.01
age2	-0.01	0.00
male	0.17	0.02
as.factor(country)Belgium	1.27	0.12
as.factor(country)Bulgaria	-1.94	0.09
as.factor(country)Canada	-1.76	0.09
as.factor(country)Czech Rep	-1.56	0.08
as.factor(country)Denmark	-1.64	0.09
as.factor(country)France	-1.71	0.10
as.factor(country)Germany	-2.11	0.09
as.factor(country)Hungary	-1.89	0.08
as.factor(country)Iceland	-1.59	0.09
as.factor(country)Ireland	-2.61	0.08
as.factor(country)Israel	-1.11	0.08
as.factor(country)Japan	-2.19	0.09
as.factor(country)Lithuania	-2.45	0.11
as.factor(country)Netherlands	-2.60	0.09
as.factor(country)New Zealand	-1.41	0.08
as.factor(country)Norway	-1.84	0.08
as.factor(country)Poland	-1.59	0.08
as.factor(country)Portugal	-1.63	0.09
as.factor(country)Slovenia	-2.93	0.09
as.factor(country)Spain	-1.84	0.08
as.factor(country)Sweden	-1.65	0.08
as.factor(country)Switzerland	-2.06	0.08
as.factor(country)UK	-1.78	0.08
as.factor(country)USA	-1.70	0.10
education:as.factor(country)Belgium	0.22	0.07
education:as.factor(country)Bulgaria	0.19	0.05
education:as.factor(country)Canada	0.21	0.05
education:as.factor(country)Czech Rep	0.06	0.05
education:as.factor(country)Denmark	0.03	0.05
education:as.factor(country)France	0.07	0.05
education:as.factor(country)Germany	0.27	0.05
education:as.factor(country)Hungary	0.21	0.05
education:as.factor(country)Iceland	0.21	0.05

education:as.factor(country)Ireland	0.06	0.05
education:as.factor(country)Israel	0.07	0.05
education:as.factor(country)Japan	-0.03	0.06
education:as.factor(country)Lithuania	0.28	0.06
education:as.factor(country)Netherlands	0.07	0.05
education:as.factor(country>New Zealand	0.05	0.04
education:as.factor(country)Norway	0.07	0.04
education:as.factor(country)Poland	0.32	0.05
education:as.factor(country)Portugal	0.12	0.05
education:as.factor(country)Slovenia	0.21	0.05
education:as.factor(country)Spain	0.14	0.05
education:as.factor(country)Sweden	0.10	0.05
education:as.factor(country)Switzerland	0.26	0.05
education:as.factor(country)UK	0.13	0.05
education:as.factor(country)USA	0.31	0.05

n = 56073, k = 53

residual deviance = 71005.8, null deviance = 77629.6 (difference = 6623.9)

The different model predictions for the effect of education on the probability of having a party ID are shown in Figure 1.

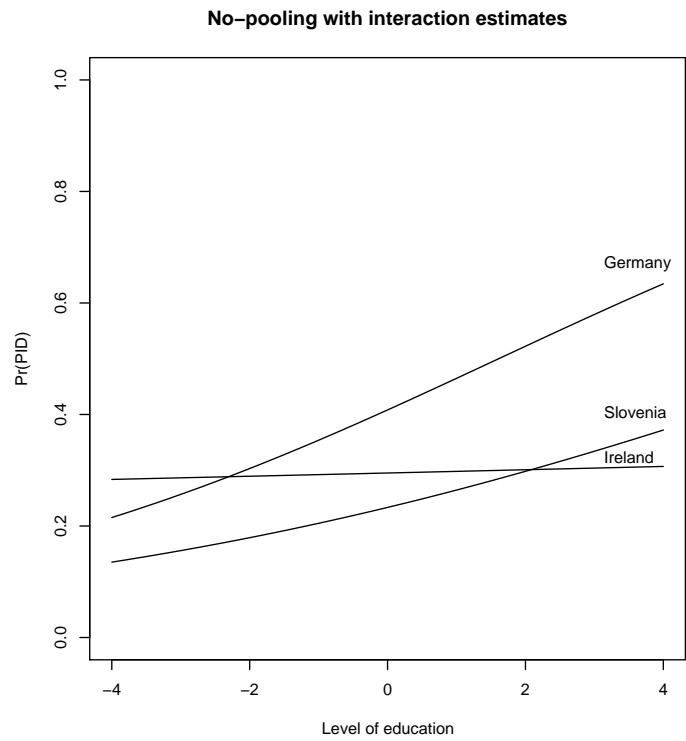
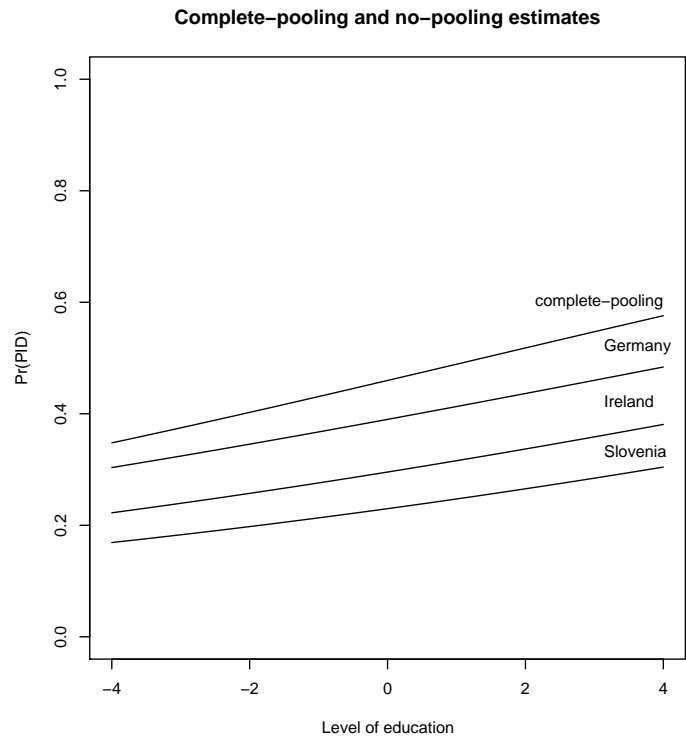


Figure 1: The effect of education on the predicted probability of identifying with a party, derived from a complete-pooling model, a no-pooling model, and a no-pooling model with interactions between the country dummy variables and education. Age variables are held at their mean values and male == 1.

d) Now repeat c with lmer

d.1) Varying intercepts model without group-level predictors

```
> lmer.d1 <- lmer(simplepid ~ education + age1 + age2 + male + (1 | country),  
+                 family=binomial(link="logit"))
```

Get coefficient estimates and standard errors for fixed effects.

Two ways: Either use `display()`:

```
> display(lmer.d1)  
lmer(formula = simplepid ~ education + age1 + age2 + male + (1 |  
      country), family = binomial(link = "logit"))  
      coef.est coef.se  
(Intercept) -0.21    0.16  
education    0.10    0.01  
age1         0.20    0.01  
age2        -0.02    0.00  
male         0.17    0.02  
  
Error terms:  
Groups   Name Std.Dev.  
country      0.81  
Residual    1.00  
---  
number of obs: 56073, groups: country, 25  
AIC = 71487.1, DIC = 71475.1  
deviance = 71475.1
```

or use `fixef()` and `se.fixef()`:

```
> fixef(lmer.d1)  
(Intercept)  education      age1      age2      male  
      -0.210      0.096      0.201      -0.016      0.172  
  
> se.fixef(lmer.d1)  
[1] 0.1623 0.0056 0.0076 0.0030 0.0179
```

Get coefficient estimates and standard errors for random effects.

Here we have to use `ranef()` and `se.ranef()` as `display()` doesn't print random effects.

```
> ranef(lmer.d1)  
An object of class "ranef.lmer"  
[[1]]  
      (Intercept)  
Australia      1.651  
Belgium        2.922  
Bulgaria       -0.255  
Canada         0.013  
Czech Rep      0.156  
Denmark        -0.021
```

France	-0.032
Germany	-0.483
Hungary	-0.236
Iceland	0.093
Ireland	-0.903
Israel	0.544
Japan	-0.500
Lithuania	-0.622
Netherlands	-0.913
New Zealand	0.268
Norway	-0.159
Poland	0.009
Portugal	0.100
Slovenia	-1.240
Spain	-0.133
Sweden	0.045
Switzerland	-0.396
UK	-0.078
USA	0.199

```
> se.ranef(lmer.d1)
An object of class "ranef.lmer"
[[1]]
      [,1]
 [1,] 0.17
 [2,] 0.19
 [3,] 0.17
 [4,] 0.17
 [5,] 0.17
 [6,] 0.17
 [7,] 0.17
 [8,] 0.17
 [9,] 0.17
[10,] 0.17
[11,] 0.17
[12,] 0.17
[13,] 0.17
[14,] 0.18
[15,] 0.17
[16,] 0.16
[17,] 0.16
[18,] 0.17
[19,] 0.17
[20,] 0.17
[21,] 0.17
[22,] 0.17
[23,] 0.17
[24,] 0.17
[25,] 0.17
```

d.2) Varying intercepts and slopes model without group-level predictors

```
> lmer.d2 <- lmer(simplepid ~ education + age1 + age2 + male + (1 + education | country),
+               family=binomial(link="logit"))
```

Get coefficient estimates and standard errors for fixed effects:

```

> fixef(lmer.d2)
(Intercept)  education      age1      age2      male
      -0.200      0.101      0.199     -0.014     0.168

> se.fixef(lmer.d2)
[1] 0.1620 0.0197 0.0077 0.0031 0.0180

```

Get coefficient estimates and standard errors for random effects:

```

> ranef(lmer.d2)
An object of class "ranef.lmer"
[[1]]
      (Intercept) education
Australia      1.6729  -0.1226
Belgium        2.9090   0.0423
Bulgaria      -0.2448   0.0455
Canada        -0.0617   0.0620
Czech Rep      0.1358  -0.0800
Denmark        0.0504  -0.1023
France        -0.0226  -0.0619
Germany       -0.4250   0.1168
Hungary       -0.1974   0.0575
Iceland        0.0992   0.0625
Ireland       -0.9149  -0.0791
Israel         0.5797  -0.0670
Japan         -0.4935  -0.1511
Lithuania     -0.7277   0.1016
Netherlands   -0.9103  -0.0678
New Zealand    0.2815  -0.0932
Norway        -0.1500  -0.0730
Poland         0.0952   0.1654
Portugal       0.0623  -0.0262
Slovenia     -1.2324   0.0599
Spain         -0.1433  -0.0032
Sweden         0.0434  -0.0378
Switzerland   -0.3662   0.1092
UK            -0.0915  -0.0104
USA           0.0091   0.1478

```

```

> se.ranef(lmer.d2)
An object of class "ranef.lmer"
[[1]]
      [,1] [,2]
[1,] 0.17 0.041
[2,] 0.19 0.052
[3,] 0.17 0.035
[4,] 0.17 0.035
[5,] 0.17 0.033
[6,] 0.17 0.031
[7,] 0.17 0.037
[8,] 0.17 0.034
[9,] 0.17 0.029
[10,] 0.17 0.032
[11,] 0.17 0.032
[12,] 0.17 0.030
[13,] 0.17 0.038

```

[14,] 0.18 0.046
[15,] 0.17 0.037
[16,] 0.16 0.025
[17,] 0.16 0.026
[18,] 0.17 0.029
[19,] 0.17 0.031
[20,] 0.17 0.036
[21,] 0.17 0.029
[22,] 0.17 0.031
[23,] 0.17 0.028
[24,] 0.17 0.033
[25,] 0.17 0.036