Simulation of Eliminating Physical Punishment With MICS Data

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Background

What would the world look like if we eliminated physical punishment? These are some quick calculations using MICS data.

Get The Data

- . clear all
- . set seed 3846
- . cd "/Users/agrogan/Desktop/newstuff/MICS-eliminate-cp"/Users/agrogan/Desktop/newstuff/MICS-eliminate-cp
- . use "/Users/agrogan/Box Sync/MICS/Data/MICS.dta"

Are we using the most up to date data?

Descriptive Statistics on Physical Punishment

. tabulate d_phys_spank

Selected child spanked	Freq.	Percent	Cum.
0	122,373	56.68	56.68
1	93,512	43.32	100.00
Total	215,885	100.00	

Predict Aggression With A Multilevel Model

For demonstration purposes, I am only including a *limited* set of covariates. One could—and should—easily include more although including more covariates considerably lengthens the estimation time.

. melogit ec16 i.d_phys_spank cmale cage || country: Fitting fixed-effects model:

```
Iteration 0:
               log\ likelihood = -142628.21
               \log \text{ likelihood} = -142431.02
Iteration 1:
Iteration 2:
               log likelihood = -142430.93
Iteration 3:
               log likelihood = -142430.93
Refining starting values:
Grid node 0:
               log likelihood = -135384.24
Fitting full model:
Iteration 0:
               log \ likelihood = -135384.24
                                              (not concave)
Iteration 1:
               log likelihood = -135381.58
                                              (backed up)
               log likelihood = -135380
Iteration 2:
                                              (backed up)
               log likelihood = -135376.47
Iteration 3:
Iteration 4:
               log\ likelihood = -135368.83
               log likelihood = -135359.89
Iteration 5:
Iteration 6:
               log\ likelihood = -135351.72
Iteration 7:
               log likelihood = -135349.08
               log \ likelihood = -135349.08
Iteration 8:
Mixed-effects logistic regression
                                                  Number of obs
                                                                          215,885
Group variable:
                         country
                                                  Number of groups
                                                                               62
                                                  Obs per group:
                                                                min =
                                                                              115
                                                                          3,482.0
                                                                 avg =
                                                                           20,451
                                                                max =
                                                  Integration pts.
Integration method: mvaghermite
                                                  Wald chi2(3)
                                                                          2481.66
Log likelihood = -135349.08
                                                  Prob > chi2
                                                                           0.0000
          ec16
                       Coef.
                               Std. Err.
                                                    P>|z|
                                                               [95% Conf. Interval]
1.d_{phys_spank}
                    .3466554
                               .0094956
                                           36.51
                                                    0.000
                                                               .3280443
                                                                           .3652665
         cmale
                    .3010048
                               .0092288
                                           32.62
                                                    0.000
                                                              .2829166
                                                                           .3190929
                  -.0060204
                                .000674
                                           -8.93
                                                    0.000
                                                             -.0073415
                                                                          -.0046993
          cage
                   -.6711418
                               .0895672
                                           -7.49
                                                    0.000
                                                             -.8466903
                                                                          -.4955932
country
                               .0778397
     var(_cons)
                    .4282121
                                                               .2998671
                                                                           .6114895
```

LR test vs. logistic model: chibar2(01) = 14163.72 Prob >= chibar2 = 0.0000

Estimate Margins (Predicted Probabilities of Aggression)

```
. margins d_phys_spank // predicted probabilities of aggression

Predictive margins Number of obs = 215,885

Model VCE : OIM

Expression : Marginal predicted mean, predict()
```

		_	Delta-method Std. Err.	z	P> z	[95% Conf.	Interval]
d_phys_s	pank						
	0	.3251464	.0169289	19.21	0.000	.2919665	.3583264
	1	.3979304	.0182745	21.78	0.000	.3621131	.4337478

Calculations

One could rely on commands such as the one below to do these calculations on-the-fly.

```
. * matrix b = r(b) // get matrix of results
.
. * matrix list b // list it out to double check
```

In this example, however, I have hand-coded the calculations, so the calculations may need to be rewritten every time more covariates are added to the model. On the other hand, writing out the calculations explicitly likely increases the transparency of the thought process below.

In a hypothetical sample of 100 children...

Aggressive Children Among Not Spanked Children

```
proportion not spanked * proportion aggressive * 100  
 . display round(.5668 * .3251464 * 100)  
 18
```

Not Aggressive Children Among Not Spanked Children

```
number\ not\ spanked-number\ aggressive
```

```
. display 57 - 18
```

Aggressive Children Among Spanked Children

```
proportion spanked * proportion aggressive * 100
    . display round(.4332 * .3979304 * 100)
    17
```

Not Aggressive Children Among Spanked Children

```
number spanked — number aggressive
. display 43 - 17
```

Number Aggressive Children Among Spanked Children If They Were Not Spanked

```
. display round(.4332 * .3251464 * 100) 14
```

Reduction in Number of Aggressive Children

```
. display 17 - 14 // this many fewer aggressive children / 100 3\,
```

Graph (DRAFT)

Is this the best graph?

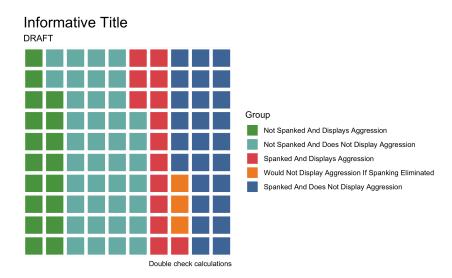


Figure 1: Graph