

# Modeling - article 1

I am going to run the models and then try to print that to a page, in landscape orientation.

First, with the raw, untransformed data. Using the data with NA removed as things like MedHHinc have some NAs.

```
mod_1 <- lmer(CRMCYPERC ~ 1 + (1|city_st), data = df_na)
mod_2 <- lmer(CRMCYPERC ~ ndvi_mean_cbg_adj + (1|city_st), data = df_na)
mod_3 <- lmer(CRMCYPERC ~ ndvi_mean_cbg_adj + (ndvi_mean_cbg_adj|city_st), data = df_na)
mod_4 <- lmer(CRMCYPERC ~ ndvi_mean_cbg_adj + MedHHinc_000 + disad + divindex + PcU18_whole + log_popden
mod_5 <- lmer(CRMCYPERC ~ ndvi_mean_cbg_adj + MedHHinc_000 + disad + divindex + PcU18_whole + log_popden
```

```
## Warning: Some predictor variables are on very different scales: consider
## rescaling
```

Landscape

```
screenreg(list(mod_1, mod_2, mod_3, mod_4, mod_5))
```

```
##
## =====
##                               Model 1      Model 2      Model 3      Model 4      Model 5
## -----
## (Intercept)                144.45 ***    304.92 ***    353.13 ***    533.03 ***    424.23 ***
##                               (5.19)      (6.85)      (13.99)      (12.50)      (14.84)
## ndvi_mean_cbg_adj          -37.23 ***    -45.79 ***    -25.94 ***    -28.39 ***
##                               (0.62)      (2.22)      (1.79)      (1.66)
## MedHHinc_000                -0.40 ***    -0.39 ***
##                               (0.02)      (0.02)
## disad                       100.22 ***    99.86 ***
##                               (0.85)      (0.85)
## divindex                    -35.00 ***    -34.81 ***
##                               (2.61)      (2.61)
## PcU18_whole                  0.04
##                               (0.05)      (0.05)
## log_popden_cbg              -29.78 ***    -29.95 ***
##                               (0.57)      (0.57)
## PerCapitaOfficers1000        4.27
##                               (3.55)
## PCgdp15                      0.00 ***
##                               (0.00)
## factor(clust90_4)2          -41.16 ***
##                               (8.93)
## factor(clust90_4)3          16.21 **
##                               (5.90)
## factor(clust90_4)4         -23.77 **
##                               (7.75)
## rate_violentcrime           15.37 ***
##                               (0.81)
## -----
## AIC                        751956.20    748461.39    744940.69    717459.74    717133.35
## BIC                        751983.19    748497.37    744994.67    717558.70    717286.28
## Log Likelihood             -375975.10  -374226.69  -372464.34  -358718.87  -358549.67
## Num. obs.                   59647      59647      59647      59647      59647
## Num. groups: city_st       301        301        301        301        301
```

```
## Var: city_st (Intercept)          7945.18      11838.47      52843.45      35386.27      18675.84
## Var: Residual                    17125.87      16114.56      15038.38      9505.99      9515.70
## Var: city_st ndvi_mean_cbg_adj          1155.13      766.89      620.43
## Cov: city_st (Intercept) ndvi_mean_cbg_adj      -6807.03      -4874.89      -3283.25
## =====
## *** p < 0.001, ** p < 0.01, * p < 0.05
```

Now with the grand centered variables

```
mod_6 <- lmer(CRMCPYPERC ~ 1 + (1|city_st), data = df_na)
mod_7 <- lmer(CRMCPYPERC ~ gndctr_ndvi_cbg_adj + (1|city_st), data = df_na)
mod_8 <- lmer(CRMCPYPERC ~ gndctr_ndvi_cbg_adj + (gndctr_ndvi_cbg_adj|city_st), data = df_na)
mod_9 <- lmer(CRMCPYPERC ~ gndctr_ndvi_cbg_adj + gndctr_MedHHinc000 + gndctr_disad + gndctr_diver + gndctr_...
mod_10 <- lmer(CRMCPYPERC ~ gndctr_ndvi_cbg_adj + gndctr_MedHHinc000 + gndctr_disad + gndctr_diver + gndctr_...
```

```
## Warning: Some predictor variables are on very different scales: consider
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```

```

##
## =====
##                               Model 1      Model 2      Model 3      Model 4      Model 5
## -----
## (Intercept)                  144.45 ***    153.62 ***    167.04 ***    166.05 ***    164.26 ***
##                               (5.19)         (6.32)         (7.55)         (5.39)         (9.61)
## gndctr_ndvi_cbg_adj          -37.23 ***    -45.79 ***    -25.94 ***    -28.39 ***
##                               (0.62)         (2.22)         (1.79)         (1.66)
## gndctr_MedHHinc000          -0.40 ***    -0.39 ***
##                               (0.02)         (0.02)
## gndctr_disad                100.22 ***    99.86 ***
##                               (0.85)         (0.85)
## gndctr_diver               -35.00 ***    -34.81 ***
##                               (2.61)         (2.61)
## gndctr_U18                   0.04         0.05
##                               (0.05)         (0.05)
## gndctr_logpopden           -29.78 ***    -29.95 ***
##                               (0.57)         (0.57)
## gndctr_Police                4.27
##                               (3.55)
## PCgdp15                      0.00 ***
##                               (0.00)
## factor(clust90_4)2          -41.16 ***
##                               (8.93)
## factor(clust90_4)3           16.21 **
##                               (5.90)
## factor(clust90_4)4          -23.77 **
##                               (7.75)
## gndctr_rt_violent           15.37 ***
##                               (0.81)
## -----
## AIC                        751956.20      748461.39      744940.69      717459.74      717133.35
## BIC                        751983.19      748497.37      744994.67      717558.70      717286.28
## Log Likelihood             -375975.10    -374226.69    -372464.34    -358718.87    -358549.67
## Num. obs.                   59647      59647      59647      59647      59647
## Num. groups: city_st        301        301        301        301        301
## Var: city_st (Intercept)     7945.18     11838.47     16593.96     8429.06     2236.71
## Var: Residual               17125.87     16114.56     15038.38     9505.99     9515.70
## Var: city_st gndctr_ndvi_cbg_adj 1155.13     766.89     620.44

```

```
## Cov: city_st (Intercept) gndctr_ndvi_cbg_adj          -2112.51      -1758.22      -761.78
## =====
## *** p < 0.001, ** p < 0.01, * p < 0.05
```

## Interaction

Using grand mean centered variables, level 1 and level 2, let's add the interaction between Green and Climate.

```
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## rescaling
```

```
##
## =====
##                                     Model 1
## -----
## (Intercept)                        162.60 ***
##                                     (10.07)
## gndctr_ndvi_cbg_adj                -27.44 ***
##                                     (3.23)
## factor(clust90_4)2                 -47.79 ***
##                                     (9.71)
## factor(clust90_4)3                  27.60 ***
##                                     (7.63)
## factor(clust90_4)4                 -25.22 **
##                                     (8.86)
## gndctr_MedHHinc000                 -0.39 ***
##                                     (0.02)
## gndctr_disad                       99.85 ***
##                                     (0.85)
## gndctr_diver                      -34.87 ***
##                                     (2.61)
## gndctr_U18                          0.05
##                                     (0.05)
## gndctr_logpopden                  -29.98 ***
##                                     (0.57)
## gndctr_Police                       3.65
##                                     (3.55)
## PCgdp15                           0.00 ***
##                                     (0.00)
## gndctr_rt_violent                  15.41 ***
##                                     (0.81)
## gndctr_ndvi_cbg_adj:factor(clust90_4)2 14.95 **
##                                     (5.25)
## gndctr_ndvi_cbg_adj:factor(clust90_4)3 -9.22 *
##                                     (4.10)
## gndctr_ndvi_cbg_adj:factor(clust90_4)4  1.94
##                                     (4.72)
## -----
## AIC                                717099.73
## BIC                                717279.66
## Log Likelihood                     -358529.87
## Num. obs.                          59647
## Num. groups: city_st                301
## Var: city_st (Intercept)            2161.43
## Var: city_st gndctr_ndvi_cbg_adj     575.58
## Cov: city_st (Intercept) gndctr_ndvi_cbg_adj -704.04
## Var: Residual                      9515.05
## =====
## *** p < 0.001, ** p < 0.01, * p < 0.05
```