

Component 2, Stage 2: SPI -> SDG

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```
knitr::opts_chunk$set(  
  echo = TRUE, warning = FALSE, message = FALSE,  
  tidy = TRUE, tidy.opts = list(width.cutoff = 60)  
)
```

1 Set up

```
# set working directory  
setwd("~/Documents/GitHub/QMSS_Thesis_Sanchez")  
  
# load libraries/packages  
source("packages.R")  
  
# load data  
source("Comp2_panel_wrangling.R")  
# select path = 'data/Main CSV'  
# Outputs/merged_cleaned_sdg.csv'  
  
# select relevant variables and arrange data  
panel_data <- panel_data %>%  
  dplyr::select(country_name, country_code, year, sdg_overall,  
    spi_comp, di_score, di_score_lag1, di_score_lag2, log_gdppc,  
    spi_comp_lag1, spi_comp_lag2, income_level_recoded) %>%  
  dplyr::arrange(country_code, year)  
  
# how many countries  
length(unique(panel_data$country_code))  
  
## [1] 162  
  
# check lag structure is correct  
head(panel_data[, c("country_code", "year", "sdg_overall", "spi_comp",  
  "spi_comp_lag1", "spi_comp_lag2", "di_score", "di_score_lag1",  
  "di_score_lag2")])  
  
## # A tibble: 6 x 9  
##   country_code year sdg_overall spi_comp spi_comp_lag1 spi_comp_lag2 di_score  
##   <chr>      <dbl>      <dbl>    <dbl>      <dbl>      <dbl>      <dbl>
```

```
## 1 AFG      2016      44.3      39.4      NA      NA      2.55
## 2 AFG      2017      45.0      44.8      39.4      NA      2.55
## 3 AFG      2018      45.5      52.0      44.8      39.4      2.97
## 4 AFG      2019      46.2      51.9      52.0      44.8      2.85
## 5 AFG      2020      47.5      55.6      51.9      52.0      2.85
## 6 AFG      2021      46.2      59.2      55.6      51.9      0.32
## # i 2 more variables: di_score_lag1 <dbl>, di_score_lag2 <dbl>
```

```
# check dimensions
dim(panel_data)
```

```
## [1] 1296    12
```

2 2.1) POLS SDG ~ SPI [Stage 2]

```
# Contemporaneous Effect: SDG ~ SPI + DI
ols_sdg_spi <- plm(formula = sdg_overall ~ spi_comp + di_score +
  log_gdppc + factor(income_level_recoded) + factor(year),
  model = "pooling", index = c("country_code", "year"), data = panel_data)
summary(ols_sdg_spi, vcov = vcovHC(ols_sdg_spi, cluster = "group",
  type = "HC1"))
```

```
## Pooling Model
```

```
##
```

```
## Note: Coefficient variance-covariance matrix supplied: vcovHC(ols_sdg_spi, cluster = "group", type =
```

```
##
```

```
## Call:
```

```
## plm(formula = sdg_overall ~ spi_comp + di_score + log_gdppc +
##   factor(income_level_recoded) + factor(year), data = panel_data,
##   model = "pooling", index = c("country_code", "year"))
```

```
##
```

```
## Unbalanced Panel: n = 155, T = 6-8, N = 1234
```

```
##
```

```
## Residuals:
```

```
##      Min.    1st Qu.      Median    3rd Qu.      Max.
## -11.96220  -2.86494   0.11405    2.84443   13.98364
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    27.473206   4.495280   6.1116 1.325e-09 ***
## spi_comp        0.284667   0.033946   8.3859 < 2.2e-16 ***
## di_score        0.291655   0.256248   1.1382 0.2552714
## log_gdppc       1.600463   0.734638   2.1786 0.0295543 *
## factor(income_level_recoded)1  4.718372   1.236004   3.8174 0.0001416 ***
## factor(income_level_recoded)2  8.481679   1.740942   4.8719 1.251e-06 ***
## factor(income_level_recoded)3  7.916414   2.577748   3.0711 0.0021802 **
## factor(year)2017    -0.276635   0.153814  -1.7985 0.0723446 .
## factor(year)2018    -0.831429   0.251746  -3.3027 0.0009855 ***
## factor(year)2019    -0.531674   0.271031  -1.9617 0.0500276 .
## factor(year)2020    -0.710819   0.363210  -1.9570 0.0505695 .
```

```
## factor(year)2021      -1.964981    0.495922 -3.9623 7.854e-05 ***
## factor(year)2022      -1.806874    0.480822 -3.7579 0.0001795 ***
## factor(year)2023      -2.226746    0.492036 -4.5256 6.612e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    124070
## Residual Sum of Squares: 24004
## R-Squared:      0.80654
## Adj. R-Squared: 0.80448
## F-statistic: 77.2465 on 13 and 154 DF, p-value: < 2.22e-16
```

```
# Adding Lag1: SPI ~ DI
```

```
ols_sdg_spi_L1 <- plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 +
  di_score + di_score_lag1 + log_gdppc + factor(income_level_recoded) +
  factor(year), model = "pooling", index = c("country_code",
  "year"), data = panel_data)
summary(ols_sdg_spi_L1, vcov = vcovHC(ols_sdg_spi_L1, cluster = "group",
  type = "HC1"))
```

```
## Pooling Model
```

```
##
```

```
## Note: Coefficient variance-covariance matrix supplied: vcovHC(ols_sdg_spi_L1, cluster = "group", type = "HC1")
```

```
##
```

```
## Call:
```

```
## plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 + di_score +
##      di_score_lag1 + log_gdppc + factor(income_level_recoded) +
##      factor(year), data = panel_data, model = "pooling", index = c("country_code",
##      "year"))
##
```

```
## Unbalanced Panel: n = 155, T = 5-7, N = 1079
```

```
##
```

```
## Residuals:
```

```
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -11.688966 -2.748015   0.076454   2.788766  14.513012
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    27.736859   4.582090   6.0533 1.964e-09 ***
## spi_comp        0.074196   0.058568   1.2668 0.205496
## spi_comp_lag1    0.214180   0.053008   4.0405 5.717e-05 ***
## di_score        0.897671   0.615320   1.4589 0.144896
## di_score_lag1   -0.651101   0.636571  -1.0228 0.306623
## log_gdppc       1.588840   0.742344   2.1403 0.032557 *
## factor(income_level_recoded)1  4.849109   1.236966   3.9202 9.415e-05 ***
## factor(income_level_recoded)2  8.484858   1.761058   4.8180 1.660e-06 ***
## factor(income_level_recoded)3  8.008641   2.641715   3.0316 0.002491 **
## factor(year)2018   -0.523275   0.158072  -3.3104 0.000963 ***
## factor(year)2019   -0.672302   0.242841  -2.7685 0.005730 **
## factor(year)2020   -0.400347   0.301988  -1.3257 0.185223
## factor(year)2021   -1.126009   0.431990  -2.6066 0.009273 **
## factor(year)2022   -2.177969   0.466997  -4.6638 3.499e-06 ***
## factor(year)2023   -2.179319   0.444974  -4.8976 1.120e-06 ***
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    107440
## Residual Sum of Squares: 20538
## R-Squared:    0.80884
## Adj. R-Squared: 0.80632
## F-statistic: 73.4697 on 14 and 154 DF, p-value: < 2.22e-16
```

```
# Adding Lag2: SPI ~ DI
```

```
ols_sdg_spi_L2 <- plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 +
  spi_comp_lag2 + di_score + di_score_lag1 + di_score_lag2 +
  log_gdppc + factor(income_level_recoded) + factor(year),
  model = "pooling", index = c("country_code", "year"), data = panel_data)
summary(ols_sdg_spi_L2, vcov = vcovHC(ols_sdg_spi_L2, cluster = "group",
  type = "HC1"))
```

```
## Pooling Model
```

```
##
```

```
## Note: Coefficient variance-covariance matrix supplied: vcovHC(ols_sdg_spi_L2, cluster = "group", type = "HC1")
```

```
##
```

```
## Call:
```

```
## plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 + spi_comp_lag2 +
##     di_score + di_score_lag1 + di_score_lag2 + log_gdppc + factor(income_level_recoded) +
##     factor(year), data = panel_data, model = "pooling", index = c("country_code",
##     "year"))
##
```

```
## Unbalanced Panel: n = 155, T = 4-6, N = 924
```

```
##
```

```
## Residuals:
```

```
##      Min.    1st Qu.      Median    3rd Qu.      Max.
## -11.71018  -2.77075    0.13186    2.83273   14.00417
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    27.587633   4.687507   5.8854 5.584e-09 ***
## spi_comp        0.091007   0.065997   1.3790 0.168245
## spi_comp_lag1    0.035694   0.032364   1.1029 0.270369
## spi_comp_lag2    0.168055   0.054651   3.0751 0.002167 **
## di_score        0.732140   0.557259   1.3138 0.189237
## di_score_lag1    0.183004   0.342883   0.5337 0.593665
## di_score_lag2   -0.722435   0.671343  -1.0761 0.282166
## log_gdppc       1.556725   0.750747   2.0736 0.038401 *
## factor(income_level_recoded)1  5.032017   1.267522   3.9700 7.756e-05 ***
## factor(income_level_recoded)2  8.604467   1.793346   4.7980 1.873e-06 ***
## factor(income_level_recoded)3  8.189760   2.711183   3.0207 0.002592 **
## factor(year)2019  -0.099031   0.203239  -0.4873 0.626188
## factor(year)2020  -0.214918   0.274614  -0.7826 0.434055
## factor(year)2021  -0.626017   0.354438  -1.7662 0.077694 .
## factor(year)2022  -1.171814   0.383213  -3.0579 0.002295 **
## factor(year)2023  -2.186161   0.447007  -4.8907 1.188e-06 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Total Sum of Squares:    91386
```

```
## Residual Sum of Squares: 17170
## R-Squared:      0.81212
## Adj. R-Squared: 0.80901
## F-statistic: 74.9353 on 15 and 154 DF, p-value: < 2.22e-16
```

2.1 Stargazer Table for POLS Models

3 2.2) First Difference [Stage 2]

```
# Contemporaneous Effect: SDG ~ SPI + DI
fd_sdg_spi <- plm(formula = sdg_overall ~ spi_comp + di_score +
  log_gdppc + factor(income_level_recoded), model = "fd", index = c("country_code",
  "year"), data = fd_data)
summary(fd_sdg_spi, vcov = vcovHC(fd_sdg_spi, cluster = "group",
  type = "HC1"))
```

```
## Oneway (individual) effect First-Difference Model
##
## Note: Coefficient variance-covariance matrix supplied: vcovHC(fd_sdg_spi, cluster = "group", type =
##
## Call:
## plm(formula = sdg_overall ~ spi_comp + di_score + log_gdppc +
##      factor(income_level_recoded), data = fd_data, model = "fd",
##      index = c("country_code", "year"))
##
## Unbalanced Panel: n = 155, T = 6-8, N = 1234
## Observations used in estimation: 1079
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -1.934635 -0.315374 -0.046495  0.248608  3.136743
##
## Coefficients:
##
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    0.3052187  0.0194557  15.6879 < 2.2e-16 ***
## spi_comp       0.0161244  0.0060174   2.6797  0.007483 **
## di_score       0.0391030  0.0844136   0.4632  0.643293
## log_gdppc     -0.2403192  0.2205508  -1.0896  0.276120
## factor(income_level_recoded)1  0.0037130  0.1825353   0.0203  0.983775
## factor(income_level_recoded)2  0.1484906  0.2066353   0.7186  0.472537
## factor(income_level_recoded)3  0.1627284  0.2219296   0.7332  0.463570
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:      348.2
## Residual Sum of Squares: 344.51
## R-Squared:      0.010603
## Adj. R-Squared: 0.0050657
## F-statistic: 1.72384 on 6 and 154 DF, p-value: 0.11891
```

```
# Adding Lag1: SPI ~ DI
```

```
fd_sdg_spi_L1 <- plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 +
  di_score + di_score_lag1 + log_gdppc + factor(income_level_recoded),
  model = "fd", index = c("country_code", "year"), data = fd_data)
summary(fd_sdg_spi_L1, vcov = vcovHC(fd_sdg_spi_L1, cluster = "group",
  type = "HC1"))
```

```
## Oneway (individual) effect First-Difference Model
```

```
##
```

```
## Note: Coefficient variance-covariance matrix supplied: vcovHC(fd_sdg_spi_L1, cluster = "group", type
```

```
##
```

```
## Call:
```

```
## plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 + di_score +
##   di_score_lag1 + log_gdppc + factor(income_level_recoded),
##   data = fd_data, model = "fd", index = c("country_code", "year"))
##
```

```
## Unbalanced Panel: n = 155, T = 5-7, N = 1079
```

```
## Observations used in estimation: 924
```

```
##
```

```
## Residuals:
```

```
##      Min.    1st Qu.      Median    3rd Qu.      Max.
## -1.929824 -0.313471 -0.045799  0.252786  3.104474
```

```
##
```

```
## Coefficients:
```

```
##
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    0.2853002  0.0250717  11.3794 < 2e-16 ***
## spi_comp       0.0115543  0.0064691   1.7861  0.07442 .
## spi_comp_lag1  0.0079235  0.0060511   1.3094  0.19071
## di_score       0.0727949  0.0843416   0.8631  0.38831
## di_score_lag1  0.1349714  0.0744059   1.8140  0.07001 .
## log_gdppc     -0.3505928  0.2483091  -1.4119  0.15831
## factor(income_level_recoded)1  0.0461826  0.2099446   0.2200  0.82594
## factor(income_level_recoded)2  0.2726074  0.2367688   1.1514  0.24988
## factor(income_level_recoded)3  0.3359719  0.2528855   1.3286  0.18433
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Total Sum of Squares:    303.64
```

```
## Residual Sum of Squares: 299.2
```

```
## R-Squared:    0.014622
```

```
## Adj. R-Squared: 0.006007
```

```
## F-statistic: 1.67189 on 8 and 154 DF, p-value: 0.10937
```

```
# Adding Lag2: SPI ~ DI
```

```
fd_sdg_spi_L2 <- plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 +
  spi_comp_lag2 + di_score + di_score_lag1 + di_score_lag2 +
  log_gdppc + factor(income_level_recoded), model = "fd", index = c("country_code",
  "year"), data = fd_data)
summary(fd_sdg_spi_L2, vcov = vcovHC(fd_sdg_spi_L2, cluster = "group",
  type = "HC1"))
```

```
## Oneway (individual) effect First-Difference Model
```

```
##
```

```
## Note: Coefficient variance-covariance matrix supplied: vcovHC(fd_sdg_spi_L2, cluster = "group", type = "HC1")
##
## Call:
## plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 + spi_comp_lag2 +
##       di_score + di_score_lag1 + di_score_lag2 + log_gdppc + factor(income_level_recoded),
##       data = fd_data, model = "fd", index = c("country_code", "year"))
##
## Unbalanced Panel: n = 155, T = 4-6, N = 924
## Observations used in estimation: 769
##
## Residuals:
##      Min.   1st Qu.   Median   3rd Qu.    Max.
## -1.85996 -0.30202 -0.03658  0.25639  2.91190
##
## Coefficients:
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    0.2339583   0.0283691   8.2469 7.159e-16 ***
## spi_comp        0.0175739   0.0070274   2.5008  0.01260 *
## spi_comp_lag1   0.0069870   0.0057285   1.2197  0.22296
## spi_comp_lag2   0.0133215   0.0059503   2.2388  0.02546 *
## di_score        0.0575724   0.0975677   0.5901  0.55532
## di_score_lag1   0.1834978   0.0765475   2.3972  0.01676 *
## di_score_lag2  -0.1765615   0.1262990  -1.3980  0.16253
## log_gdppc       -0.4945605   0.2413981  -2.0487  0.04083 *
## factor(income_level_recoded)1  0.1177717   0.1735894   0.6784  0.49769
## factor(income_level_recoded)2  0.3190698   0.2020118   1.5795  0.11465
## factor(income_level_recoded)3  0.3913713   0.2137640   1.8309  0.06751 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    229.74
## Residual Sum of Squares: 221.66
## R-Squared:    0.035145
## Adj. R-Squared: 0.022416
## F-statistic: 2.16212 on 10 and 154 DF, p-value: 0.022917
```

3.1 Stargazer Table for FD Models

4 2.3) Fixed Effects [Stage 2]

```
# Contemporaneous Effect: SDG ~ SPI + DI
fe_sdg_spi <- plm(formula = sdg_overall ~ spi_comp + di_score +
  log_gdppc + factor(year) + factor(income_level_recoded),
  index = c("country_code", "year"), data = panel_data, model = "within" #FE
)
summary(fe_sdg_spi, vcov = vcovHC(fe_sdg_spi, cluster = "group",
  type = "HC1")) # Robust SEs
```

```
## Oneway (individual) effect Within Model
##
```

```
## Note: Coefficient variance-covariance matrix supplied: vcovHC(fe_sdg_spi, cluster = "group", type = "HC1")
```

```
##
## Call:
## plm(formula = sdg_overall ~ spi_comp + di_score + log_gdppc +
##       factor(year) + factor(income_level_recoded), data = panel_data,
##       model = "within", index = c("country_code", "year"))
##
## Unbalanced Panel: n = 155, T = 6-8, N = 1234
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -2.8369120 -0.3092824 -0.0081391  0.3334427  3.4434607
##
## Coefficients:
##
##              Estimate Std. Error t-value Pr(>|t|)
## spi_comp          0.044006   0.014857  2.9619  0.003126 **
## di_score          0.117373   0.089489  1.3116  0.189938
## log_gdppc        0.337323   0.315447  1.0693  0.285154
## factor(year)2017   0.342538   0.053932  6.3512 3.157e-10 ***
## factor(year)2018   0.554374   0.092605  5.9865 2.927e-09 ***
## factor(year)2019   0.946068   0.098688  9.5865 < 2.2e-16 ***
## factor(year)2020   1.258547   0.108831 11.5642 < 2.2e-16 ***
## factor(year)2021   1.248780   0.164380  7.5969 6.617e-14 ***
## factor(year)2022   1.445105   0.166247  8.6925 < 2.2e-16 ***
## factor(year)2023   1.616792   0.182979  8.8359 < 2.2e-16 ***
## factor(income_level_recoded)1 0.391704   0.398058  0.9840 0.325320
## factor(income_level_recoded)2 0.738052   0.475931  1.5508 0.121258
## factor(income_level_recoded)3 0.371058   0.545743  0.6799 0.496707
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    1162.9
## Residual Sum of Squares: 441.55
## R-Squared:      0.6203
## Adj. R-Squared: 0.56082
## F-statistic: 45.9504 on 13 and 154 DF, p-value: < 2.22e-16
```

Adding Lag1: SPI ~ DI

```
fe_sdg_spi_L1 <- plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 +
  di_score + di_score_lag1 + log_gdppc + factor(income_level_recoded) +
  factor(year), model = "within", index = c("country_code",
  "year"), data = panel_data)
summary(fe_sdg_spi_L1, vcov = vcovHC(fe_sdg_spi_L1, cluster = "group",
  type = "HC1"))
```

```
## Oneway (individual) effect Within Model
```

```
##
```

```
## Note: Coefficient variance-covariance matrix supplied: vcovHC(fe_sdg_spi_L1, cluster = "group", type
```

```
##
```

```
## Call:
```

```
## plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 + di_score +
##       di_score_lag1 + log_gdppc + factor(income_level_recoded) +
##       factor(year), data = panel_data, model = "within", index = c("country_code",
##       "year"))
##
```



```
## Unbalanced Panel: n = 155, T = 5-7, N = 1079
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -2.658946 -0.271198 -0.013632  0.271782  3.140399
##
## Coefficients:
##              Estimate Std. Error t-value Pr(>|t|)
## spi_comp          0.0241741  0.0112967  2.1399  0.03263 *
## spi_comp_lag1      0.0237987  0.0096619  2.4632  0.01396 *
## di_score           0.2374865  0.0942257  2.5204  0.01189 *
## di_score_lag1     -0.0458757  0.1096028 -0.4186  0.67563
## log_gdppc          0.3339167  0.3676600  0.9082  0.36400
## factor(income_level_recoded)1 0.3728521  0.3144906  1.1856  0.23610
## factor(income_level_recoded)2 0.7585034  0.4086083  1.8563  0.06373 .
## factor(income_level_recoded)3 0.3845908  0.5307062  0.7247  0.46884
## factor(year)2018      0.2049135  0.0641109  3.1962  0.00144 **
## factor(year)2019      0.5479565  0.0781320  7.0132 4.541e-12 ***
## factor(year)2020      0.9074314  0.0921364  9.8488 < 2.2e-16 ***
## factor(year)2021      0.9466398  0.1360859  6.9562 6.676e-12 ***
## factor(year)2022      1.0123227  0.1617134  6.2600 5.925e-10 ***
## factor(year)2023      1.2283736  0.1690011  7.2684 7.826e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:      740.6
## Residual Sum of Squares: 312.19
## R-Squared:      0.57847
## Adj. R-Squared: 0.50065
## F-statistic: 33.8593 on 14 and 154 DF, p-value: < 2.22e-16
```

```
# Adding Lag2: SPI ~ DI
fe_sdg_spi_L2 <- plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 +
  spi_comp_lag2 + di_score + di_score_lag1 + di_score_lag2 +
  log_gdppc + factor(income_level_recoded) + factor(year),
  model = "within", index = c("country_code", "year"), data = panel_data)
summary(fe_sdg_spi_L2, vcov = vcovHC(fe_sdg_spi_L2, cluster = "group",
  type = "HC1"))
```

```
## Oneway (individual) effect Within Model
##
## Note: Coefficient variance-covariance matrix supplied: vcovHC(fe_sdg_spi_L2, cluster = "group", type
##
## Call:
## plm(formula = sdg_overall ~ spi_comp + spi_comp_lag1 + spi_comp_lag2 +
##      di_score + di_score_lag1 + di_score_lag2 + log_gdppc + factor(income_level_recoded) +
##      factor(year), data = panel_data, model = "within", index = c("country_code",
##      "year"))
##
## Unbalanced Panel: n = 155, T = 4-6, N = 924
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -2.4105174 -0.2462874 -0.0016371  0.2405397  2.8719407
```

```
##
## Coefficients:
##               Estimate Std. Error t-value Pr(>|t|)
## spi_comp        0.0338896  0.0117720  2.8788 0.004105 **
## spi_comp_lag1    0.0063699  0.0075753  0.8409 0.400685
## spi_comp_lag2    0.0163276  0.0085846  1.9020 0.057557 .
## di_score         0.1559032  0.0825031  1.8897 0.059186 .
## di_score_lag1    0.1701864  0.0971166  1.7524 0.080113 .
## di_score_lag2   -0.2460409  0.1258641 -1.9548 0.050974 .
## log_gdppc       -0.0725105  0.3338289 -0.2172 0.828105
## factor(income_level_recoded)1 0.4185389  0.1920359  2.1795 0.029605 *
## factor(income_level_recoded)2 0.7948941  0.2748999  2.8916 0.003943 **
## factor(income_level_recoded)3 0.5141073  0.3689604  1.3934 0.163911
## factor(year)2019  0.3334744  0.0438533  7.6043 8.534e-14 ***
## factor(year)2020  0.6107261  0.0763721  7.9967 4.803e-15 ***
## factor(year)2021  0.6878165  0.1068534  6.4370 2.167e-10 ***
## factor(year)2022  0.8321507  0.1292528  6.4382 2.151e-10 ***
## factor(year)2023  0.9354750  0.1628894  5.7430 1.349e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    440.52
## Residual Sum of Squares: 198.57
## R-Squared:    0.54924
## Adj. R-Squared: 0.44821
## F-statistic: 29.6661 on 15 and 154 DF, p-value: < 2.22e-16
```

5 Stargazer Table for FE Models

5.1 make a stargazer table of all Lag2 models for POLS, FD and FE

5.2 Fixed Effects Error Bar Visualization

```
# Extract coefficients and robust standard errors from the
# FE model
coefs <- summary(fe_sdg_spi_L2, vcov = vcovHC(fe_sdg_spi_L2,
  cluster = "group", type = "HC1"))$coefficients
# Create a data frame for visualization
coef_df <- data.frame(term = rownames(coefs), estimate = coefs[,
  "Estimate"], std.error = coefs[, "Std. Error"])
# Create a ggplot error bar chart
ebar_fe <- ggplot(coef_df, aes(x = term, y = estimate)) + geom_point() +
  geom_errorbar(aes(ymin = estimate - std.error, ymax = estimate +
    std.error), width = 0.2) + labs(title = "Stage II: FE Model (sdg_overall ~ spi_comp + di_score)
  x = NULL, y = "Coefficient Estimate") + theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  scale_x_discrete(labels = c(di_score = "Democracy (DI)",
    di_score_lag1 = "Lagged DI (t-1)", di_score_lag2 = "Lagged DI (t-2)",
    spi_comp = "Statistical Performance (SPI)", spi_comp_lag1 = "Lagged SPI (t-1)",
    spi_comp_lag2 = "Lagged SPI (t-2)", log_gdppc = "Log(GDP per capita)",
    `factor(income_level_recoded)1` = "Lower-Middle Income (LMIC)",
```

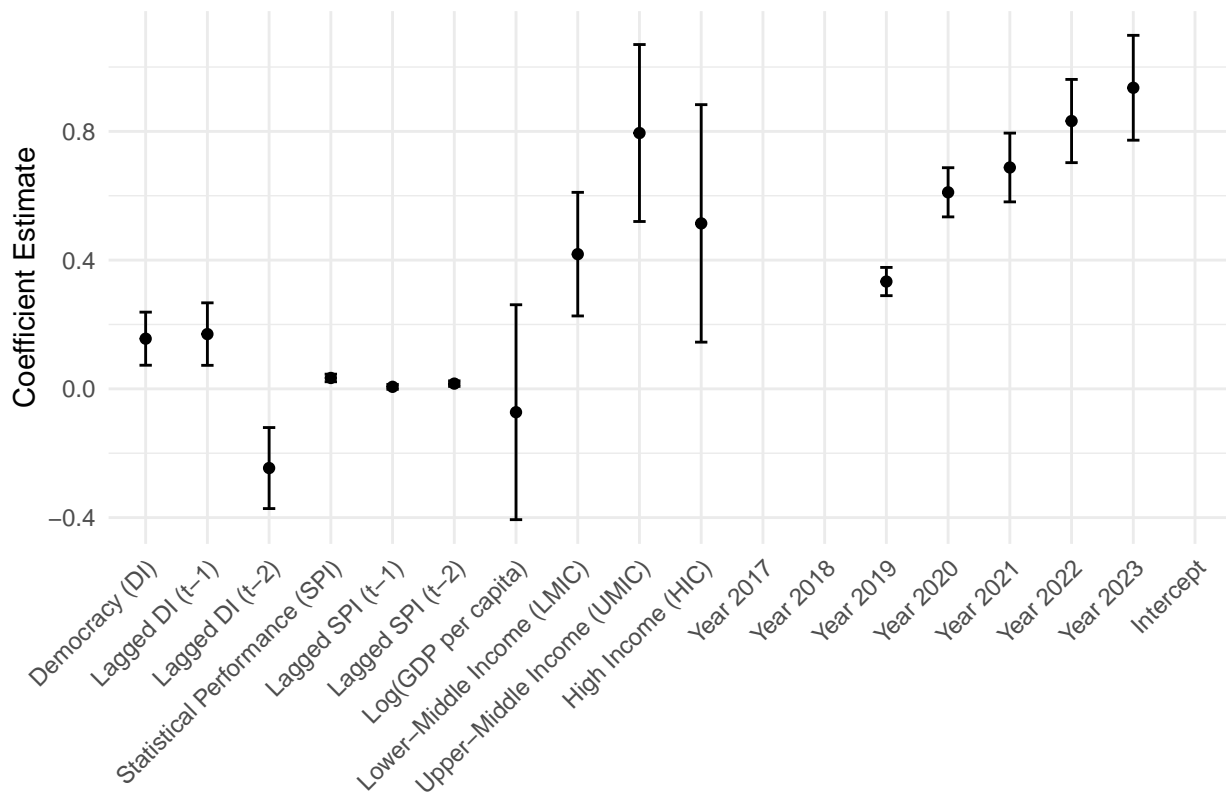
```

`factor(income_level_recoded)2` = "Upper-Middle Income (UMIC)",
`factor(income_level_recoded)3` = "High Income (HIC)",
`factor(year)2017` = "Year 2017", `factor(year)2018` = "Year 2018",
`factor(year)2019` = "Year 2019", `factor(year)2020` = "Year 2020",
`factor(year)2021` = "Year 2021", `factor(year)2022` = "Year 2022",
`factor(year)2023` = "Year 2023", Intercept = "Intercept"),
limits = c("di_score", "di_score_lag1", "di_score_lag2",
  "spi_comp", "spi_comp_lag1", "spi_comp_lag2", "log_gdppc",
  "factor(income_level_recoded)1", "factor(income_level_recoded)2",
  "factor(income_level_recoded)3", "factor(year)2017",
  "factor(year)2018", "factor(year)2019", "factor(year)2020",
  "factor(year)2021", "factor(year)2022", "factor(year)2023",
  "Intercept"))

```

ebar_fe

Stage II: FE Model (sdg_overall ~ spi_comp + di_score)



```

# Save the plot
# ggsave('component_2/figures/stage2/error_bar_fe_sdg_spi_L2.png',
# ebar_fe, width = 10, height = 6)

```

5.3 First Difference Error Bar Visualization

```

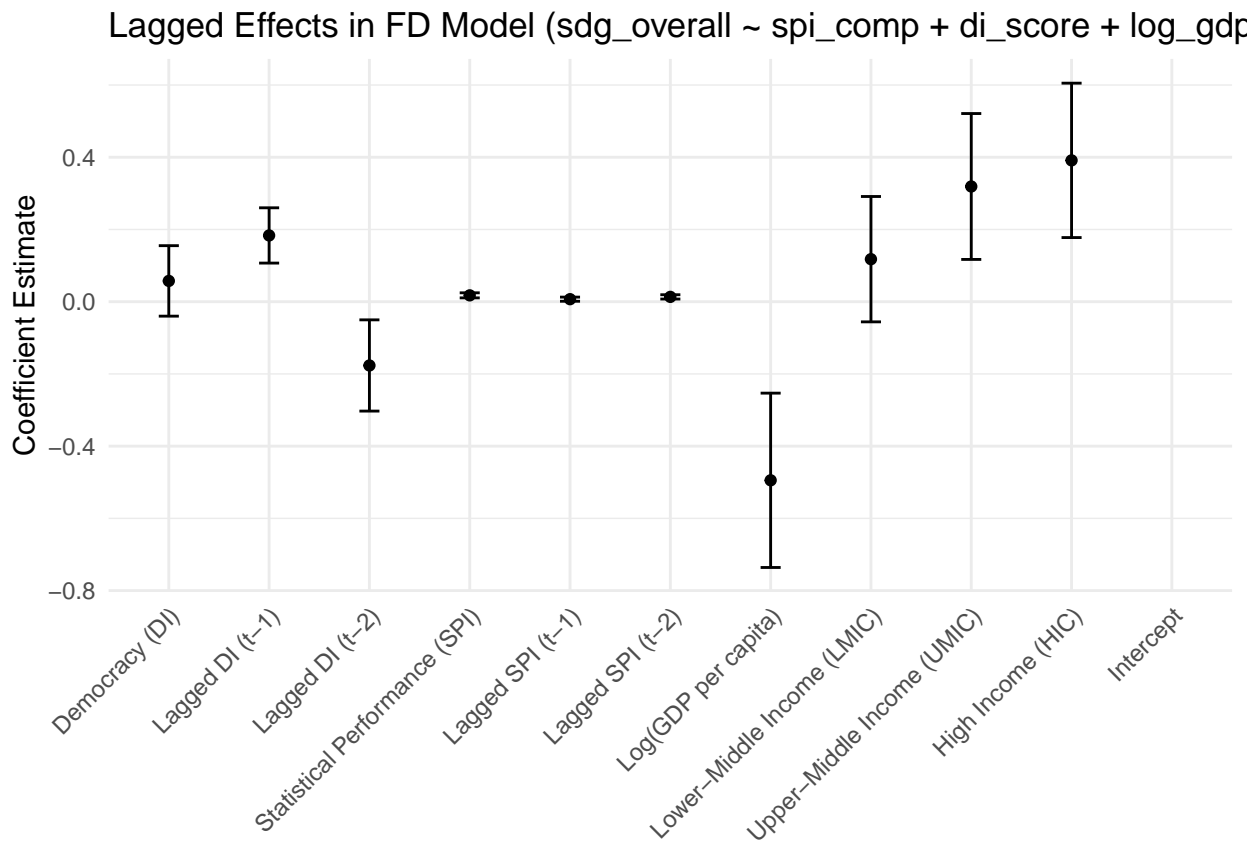
# Extract coefficients and robust standard errors from the
# FD model

```

```

coefs_fd <- summary(fd_sdg_spi_L2, vcov = vcovHC(fd_sdg_spi_L2,
  cluster = "group", type = "HC1"))$coefficients
# Create a data frame for visualization
coef_df_fd <- data.frame(term = rownames(coefs_fd), estimate = coefs_fd[,
  "Estimate"], std.error = coefs_fd[, "Std. Error"])
# Create a ggplot error bar chart for the FD model
ebar_fd <- ggplot(coef_df_fd, aes(x = term, y = estimate)) +
  geom_point() + geom_errorbar(aes(ymin = estimate - std.error,
  ymax = estimate + std.error), width = 0.2) + labs(title = "Lagged Effects in FD Model (sdg_overall ~ spi_comp + di_score + log_gdppc)",
  x = NULL, y = "Coefficient Estimate") + theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  scale_x_discrete(labels = c(di_score = "Democracy (DI)",
    di_score_lag1 = "Lagged DI (t-1)", di_score_lag2 = "Lagged DI (t-2)",
    spi_comp = "Statistical Performance (SPI)", spi_comp_lag1 = "Lagged SPI (t-1)",
    spi_comp_lag2 = "Lagged SPI (t-2)", log_gdppc = "Log(GDP per capita)",
    `factor(income_level_recoded)1` = "Lower-Middle Income (LMIC)",
    `factor(income_level_recoded)2` = "Upper-Middle Income (UMIC)",
    `factor(income_level_recoded)3` = "High Income (HIC)",
    Intercept = "Intercept"), limits = c("di_score", "di_score_lag1",
    "di_score_lag2", "spi_comp", "spi_comp_lag1", "spi_comp_lag2",
    "log_gdppc", "factor(income_level_recoded)1", "factor(income_level_recoded)2",
    "factor(income_level_recoded)3", "Intercept"))
ebar_fd

```



```
# Save the plot
# ggsave('component_2/figures/stage2/error_bar_fd_sdg_spi_L2.png',
# ebar_fd, width = 10, height = 6)
```

5.4 Check for Autocorrelation

```
# APPLY Wooldridge Test for AR(1) Errors in FE Panel
# Models: pwartest()
# https://search.r-project.org/CRAN/refmans/plm/html/pwartest.html
# This is MUCH BETTER for panel data with small T AND
# unbalanced panels!!!
pwartest(fe_sdg_spi_L2) # [significant]
```

```
##
## Wooldridge's test for serial correlation in FE panels
##
## data: fe_sdg_spi_L2
## F = 67.496, df1 = 1, df2 = 767, p-value = 8.955e-16
## alternative hypothesis: serial correlation
```

Significant p-value indicates the presence of autocorrelation in the residuals of the fixed effects model. This suggests that the errors are correlated over time, which violates one of the key assumptions of linear regression models.

5.5 Check for Heteroskedasticity

```
# Apply Breusch-Pagan test for heteroskedasticity
bptest(fe_sdg_spi_L2, studentize = TRUE) # Heteroskedasticity [significant]
```

```
##
## studentized Breusch-Pagan test
##
## data: fe_sdg_spi_L2
## BP = 98.322, df = 15, p-value = 2.711e-14
```

```
bptest(fd_sdg_spi_L2, studentize = TRUE) # Heteroskedasticity [significant]
```

```
##
## studentized Breusch-Pagan test
##
## data: fd_sdg_spi_L2
## BP = 93.383, df = 10, p-value = 1.14e-15
```

The Breusch-Pagan test indicates the presence of heteroskedasticity in the residuals of the fixed effects model. This suggests that the variance of the errors is not constant across observations, which violates another key assumption of linear regression models.

Both violations are corrected by using robust standard errors clustered at the country level, which accounts for autocorrelation and heteroskedasticity.