

modelo_tfm

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Pre procesamiento de datos

1. Lectura de datos y formato panel

```
df <- read.csv('df_17_21_noclean.csv') %>%
  dplyr::select(year, country, total_earnings,
               total_players, pbicap, desempleo,
               inflacion, internet, poblacion, pea,
               net_mig) %>%
  arrange(country, decreasing = FALSE)

# rango de fechas
df <- df %>%
  filter(year < as.Date("2022-01-01"))

# Filtro NA
pbicap_faltantes <- unique(df[is.na(df$pbicap), ]$country) #"Cuba" "Venezuela"
df <- df[!df$country %in% pbicap_faltantes, ]

df <- df[!df$country %in% c("Cambodia","Syrian Arab Republic","Trinidad and Tobago",
                           "Mongolia","Hong Kong"), ]
df[df$country=='Korea, Republic of', 'net_mig'][4] <- 32000 #duda porque en otra web es 0

# renombrar variables para el modelo
renombrar <- c("year", "pais", "TG", "TJ", "PIBc",
               "DEmp", "INF", "INT", "POB", "PEA", "MIG")

colnames(df) <- renombrar
```

Revision de nullos

```
sapply(df, function(x) sum(is.na(x)))
```

```
## year pais   TG   TJ PIBc DEmp  INF  INT  POB  PEA  MIG
##    0    0    0    0   0    0    0   0   0   0   0
```

Revision de total de datos

```
df %>%  
  group_by(year) %>%  
  summarise(count = n())
```

```
## # A tibble: 5 x 2  
##   year      count  
##   <chr>    <int>  
## 1 2017-01-01     92  
## 2 2018-01-01     92  
## 3 2019-01-01     92  
## 4 2020-01-01     92  
## 5 2021-01-01     92
```

Descriptivo

```
# sumtable(df[, -c(1,2)], digits = 3,  
#          title="Estadísticos descriptivos")
```

```
# nuevas variables  
df$TJ_POB <- (df$TJ/df$POB)  
df$PEA_POB <- (df$PEA/df$POB)  
df$MIG_POB <- (df$MIG/df$POB)
```

```
summary(df[, -c(1,2)])
```

```
##          TG          TJ          PIBc          DEmp  
## Min.   :    55   Min.   :   1.0   Min.   : 1243   Min.   : 0.720  
## 1st Qu.: 57244   1st Qu.: 20.0   1st Qu.: 4763   1st Qu.: 3.917  
## Median : 289862   Median : 80.5   Median : 12301   Median : 5.555  
## Mean   : 1954964   Mean   : 278.2   Mean   : 22911   Mean   : 7.070  
## 3rd Qu.: 1398620   3rd Qu.: 293.2   3rd Qu.: 36706   3rd Qu.: 9.037  
## Max.   :51549194   Max.   :6279.0   Max.   :133712   Max.   :28.770  
##          INF          INT          POB          PEA  
## Min.   : -13.911   Min.   : 13.78   Min.   :3.434e+05   Min.   : 228127  
## 1st Qu.:  1.549   1st Qu.: 65.36   1st Qu.:5.432e+06   1st Qu.: 3496121  
## Median :  2.893   Median : 79.75   Median :1.126e+07   Median : 7375306  
## Mean   :  5.382   Mean   : 75.49   Mean   :6.831e+07   Mean   : 45670295  
## 3rd Qu.:  5.380   3rd Qu.: 88.93   3rd Qu.:4.765e+07   3rd Qu.: 31804312  
## Max.   :150.001   Max.   :100.00   Max.   :1.412e+09   Max.   :986411437  
##          MIG          TJ_POB          PEA_POB          MIG_POB  
## Min.   : -1574581   Min.   :6.040e-09   Min.   :0.5718   Min.   : -0.0582915  
## 1st Qu.: -16753   1st Qu.:1.616e-06   1st Qu.:0.6409   1st Qu.: -0.0013823  
## Median :  3094   Median :5.845e-06   Median :0.6558   Median : 0.0001741  
## Mean   :  15616   Mean   :1.274e-05   Mean   :0.6624   Mean   : 0.0008615  
## 3rd Qu.:  35468   3rd Qu.:1.543e-05   3rd Qu.:0.6793   3rd Qu.: 0.0034413  
## Max.   : 1377630   Max.   :1.276e-04   Max.   :0.8460   Max.   : 0.0341062
```

3. Normalizacion con logaritmo

- valores con varianzas muy grandes
- aplico normalizacion logaritmica en algunas variables

```
df_standar <- df %>%  
  mutate(across(c("TG", "PIBc", "POB", "PEA", "TJ", "TJ_POB", "PEA_POB"), ~log(.)))%>%  
  mutate(year = year(df$year))  
  
summary(df_standar)
```

```
##      year      pais      TG      TJ  
## Min.   :2017   Length:460   Min.   : 4.006   Min.   :0.000  
## 1st Qu.:2018   Class :character   1st Qu.:10.955   1st Qu.:2.996  
## Median :2019   Mode  :character   Median :12.577   Median :4.388  
## Mean   :2019                                     Mean   :12.471   Mean   :4.317  
## 3rd Qu.:2020                                     3rd Qu.:14.151   3rd Qu.:5.681  
## Max.   :2021                                     Max.   :17.758   Max.   :8.745  
##      PIBc      DEmp      INF      INT  
## Min.   : 7.125   Min.   : 0.720   Min.   : -13.911   Min.   : 13.78  
## 1st Qu.: 8.469   1st Qu.: 3.917   1st Qu.:  1.549   1st Qu.: 65.36  
## Median : 9.417   Median : 5.555   Median :  2.893   Median : 79.75  
## Mean   : 9.456   Mean   : 7.070   Mean   :  5.382   Mean   : 75.49  
## 3rd Qu.:10.511   3rd Qu.: 9.037   3rd Qu.:  5.380   3rd Qu.: 88.93  
## Max.   :11.803   Max.   :28.770   Max.   :150.001   Max.   :100.00  
##      POB      PEA      MIG      TJ_POB  
## Min.   :12.75   Min.   :12.34   Min.   : -1574581   Min.   : -18.925  
## 1st Qu.:15.51   1st Qu.:15.07   1st Qu.:  -16753   1st Qu.: -13.335  
## Median :16.24   Median :15.81   Median :   3094   Median : -12.050  
## Mean   :16.58   Mean   :16.16   Mean   :  15616   Mean   : -12.258  
## 3rd Qu.:17.68   3rd Qu.:17.27   3rd Qu.:  35468   3rd Qu.: -11.079  
## Max.   :21.07   Max.   :20.71   Max.   : 1377630   Max.   :  -8.967  
##      PEA_POB      MIG_POB  
## Min.   : -0.5590   Min.   : -0.0582915  
## 1st Qu.: -0.4448   1st Qu.: -0.0013823  
## Median : -0.4219   Median :  0.0001741  
## Mean   : -0.4135   Mean   :  0.0008615  
## 3rd Qu.: -0.3866   3rd Qu.:  0.0034413  
## Max.   : -0.1673   Max.   :  0.0341062
```

0. Preparando los datos

- Tenemos datos panel con la siguiente forma 90 paises 5 años y estas columnas
- Nuestro panel es balanceado y corto

```
dim(table(df_standar$pais,df_standar$year))
```

```
## [1] 92  5
```

```
colnames(df_standar)
```

```
## [1] "year"      "pais"      "TG"        "TJ"        "PIBc"      "DEmp"      "INF"
## [8] "INT"       "POB"       "PEA"       "MIG"       "TJ_POB"    "PEA_POB"   "MIG_POB"
```

- definimos las variables para el modelo

```
attach(df_standar)
Y <- cbind(TG)
X <- cbind(PIBc, INT, DEmp, INF, TJ_POB, PEA_POB, MIG_POB)

df_panel <- pdata.frame(df_standar,
                        index=c('pais','year'))

# para las validaciones
#####
datos_2 = data.frame(Y, X, pais)
write.table(datos_2, "02_datos.txt", sep=";", row.names=F)
#####

head(df_panel,3)
```

```
##           year   pais      TG      TJ      PIBc  DEmp      INF      INT
## Albania-2017 2017 Albania  7.961426 0.6931472 8.418705 13.62 1.450732 62.40000
## Albania-2018 2018 Albania  7.205301 1.0986123 8.573131 12.30 1.472953 65.40000
## Albania-2019 2019 Albania 10.531019 2.6390573 8.593453 11.47 1.257025 68.55039
##                POB      PEA  MIG    TJ_POB    PEA_POB      MIG_POB
## Albania-2017 14.87103 14.48765 -9768 -14.17788 -0.3833763 -0.003399390
## Albania-2018 14.86856 14.48388 -9106 -13.76995 -0.3846839 -0.003176834
## Albania-2019 14.86430 14.47713 -8889 -12.22524 -0.3871680 -0.003114368
```

1. Efectos Fijos

```
fijos <- plm(Y ~ X, data=df_panel, index=c('pais','year'), model= "within")
# summary(fijos)
```

2. Efectos aleatorios

```
random <- plm(Y ~ X, data=df_panel, index=c('pais','year'), model= "random")
# summary(random)
```

3. MCO

```
mco_pool = plm(Y ~ X, data=df_panel, index=c("state", "year"), model="pooling")
mco = lm(Y ~ X, data=df_panel)
# summary(mco)
```

Test para escoger el mejor modelo

1. F test

- H_0 : modelo (MCO) vs H_1 : efectos fijos
- $p < 0.05$ entonces rechazo H_0 , el mejor modelo seria efectos fijos

```
pFtest(fijos, mco)
```

```
##  
## F test for individual effects  
##  
## data: Y ~ X  
## F = 32.956, df1 = 91, df2 = 361, p-value < 2.2e-16  
## alternative hypothesis: significant effects
```

2. Breusch-Pagan

- H_0 : modelo agrupado (MCO) vs H_1 : efectos aleatorios
- $p < 0.05$ entonces rechazo la H_0 , por ahora el mejor modelo seria aleatorios

```
plmtest(mco_pool, type=c("bp"))
```

```
##  
## Lagrange Multiplier Test - (Breusch-Pagan)  
##  
## data: Y ~ X  
## chisq = 649.43, df = 1, p-value < 2.2e-16  
## alternative hypothesis: significant effects
```

3. Hausman test

- H_0 : efectos aleatorios vs H_1 : efectos fijos
- $p < 0.05$ entonces rechazo H_0 y decido que efectos fijos es mejor

```
phtest(fijos, random)
```

```
##  
## Hausman Test  
##  
## data: Y ~ X  
## chisq = 39.202, df = 7, p-value = 1.789e-06  
## alternative hypothesis: one model is inconsistent
```

Regresiones

Regresion con efectos fijos

by pais Spain

```
df_panel$pais <- relevel(df_panel$pais, ref = "Spain")
regression_pais_sp = lm(Y ~ X + factor(pais), data = df_panel)

# summary(regression_pais)

p_values <- summary(regression_pais_sp)$coefficients[,4]
coeficiente <- summary(regression_pais_sp)$coefficients[,1]

no_significativo <- names(p_values)[which(p_values > 0.05)]

significativo_positivos <- names(p_values)[which(p_values < 0.05 & coeficiente>0)]
significativo_negativos <- names(p_values)[which(p_values < 0.05 & coeficiente<0)]
```

```
summary(regression_pais_sp)
```

```
##
## Call:
## lm(formula = Y ~ X + factor(pais), data = df_panel)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.76996 -0.36105 -0.02916  0.32902  2.84409
##
## Coefficients:
##              Estimate Std. Error t value
## (Intercept)    21.06503    4.823295   4.367
## XPIBc           0.769572   0.411327   1.871
## XINT            0.035852   0.007456   4.808
## XDEmp          -0.065721   0.034901  -1.883
## XINF            0.004736   0.005246   0.903
## XTJ_POB         1.025040   0.084437  12.140
## XPEA_POB        12.473900   5.503209   2.267
## XMIG_POB        -7.188313   8.081750  -0.889
## factor(pais)Albania    -2.144663   0.843753  -2.542
## factor(pais)Algeria     2.618401   0.985977   2.656
## factor(pais)Argentina    0.831639   0.683290   1.217
## factor(pais)Armenia    -0.811489   0.904649  -0.897
## factor(pais)Australia   -1.317981   0.597802  -2.205
## factor(pais)Austria     -2.706809   0.604721  -4.476
## factor(pais)Azerbaijan  -2.017365   1.041847  -1.936
## factor(pais)Bahrain     -6.018967   1.196906  -5.029
## factor(pais)Bangladesh    2.654461   1.271201   2.088
## factor(pais)Belarus     -0.718528   0.917352  -0.783
## factor(pais)Belgium     -1.907331   0.574777  -3.318
```

## factor(pais)Bolivia	1.862281	1.149941	1.619
## factor(pais)Bosnia and Herzegovina	0.506333	0.762587	0.664
## factor(pais)Brazil	2.637609	0.745351	3.539
## factor(pais)Bulgaria	0.896528	0.796969	1.125
## factor(pais)Canada	-0.603843	0.566714	-1.066
## factor(pais)Chile	-1.693578	0.672409	-2.519
## factor(pais)China	5.436857	0.836287	6.501
## factor(pais)Colombia	0.852314	0.830547	1.026
## factor(pais)Costa Rica	-3.279044	0.653329	-5.019
## factor(pais)Croatia	-1.267382	0.633345	-2.001
## factor(pais)Czech Republic	-1.726594	0.678200	-2.546
## factor(pais)Denmark	-1.634979	0.629256	-2.598
## factor(pais)Dominican Republic	-0.073878	0.828520	-0.089
## factor(pais)Ecuador	0.084063	0.937223	0.090
## factor(pais)Egypt	3.465728	1.148155	3.019
## factor(pais)Estonia	-2.447960	0.631293	-3.878
## factor(pais)Finland	-1.067171	0.631063	-1.691
## factor(pais)France	1.305914	0.620434	2.105
## factor(pais)Georgia	-1.731973	0.903558	-1.917
## factor(pais)Germany	0.171319	0.614796	0.279
## factor(pais)Greece	-0.500276	0.553675	-0.904
## factor(pais)Guatemala	1.164119	1.142178	1.019
## factor(pais)Hungary	-2.144792	0.691082	-3.104
## factor(pais)Iceland	-7.675532	0.670868	-11.441
## factor(pais)India	6.167992	1.237244	4.985
## factor(pais)Indonesia	3.705709	1.059920	3.496
## factor(pais)Iran, Islamic Republic of	0.310060	1.007559	0.308
## factor(pais)Iraq	4.279204	1.192420	3.589
## factor(pais)Ireland	-3.495248	0.654849	-5.337
## factor(pais)Israel	-0.296378	0.734765	-0.403
## factor(pais)Italy	0.889506	0.538354	1.652
## factor(pais)Japan	2.014307	0.833840	2.416
## factor(pais)Jordan	4.286066	0.959388	4.467
## factor(pais)Kazakhstan	0.862337	0.814455	1.059
## factor(pais)Korea, Republic of	-0.813170	0.861786	-0.944
## factor(pais)Kuwait	-5.721402	1.071966	-5.337
## factor(pais)Kyrgyzstan	1.811739	1.521138	1.191
## factor(pais)Lao People's Democratic Republic	-0.346902	1.245702	-0.278
## factor(pais)Latvia	-3.090577	0.650984	-4.748
## factor(pais)Lebanon	0.958431	0.846782	1.132
## factor(pais)Lithuania	-3.029636	0.614659	-4.929
## factor(pais)Luxembourg	-8.342336	0.838396	-9.950
## factor(pais)Macao	-5.553994	1.042911	-5.325
## factor(pais)Malaysia	-0.130647	0.863736	-0.151
## factor(pais)Malta	-7.343612	0.671328	-10.939
## factor(pais)Mexico	2.190249	0.803229	2.727
## factor(pais)Moldova, Republic of	-0.760414	1.106059	-0.687
## factor(pais)Morocco	1.173170	1.037621	1.131
## factor(pais)Netherlands	-1.759174	0.617435	-2.849
## factor(pais>New Zealand	-3.011032	0.595430	-5.057
## factor(pais>Nicaragua	1.671960	1.296453	1.290
## factor(pais)North Macedonia	-1.177932	0.783688	-1.503
## factor(pais>Norway	-3.134177	0.677036	-4.629
## factor(pais)Pakistan	8.035436	1.604058	5.009

## factor(pais)Panama	-2.361322	0.624309	-3.782
## factor(pais)Paraguay	-2.447527	0.916895	-2.669
## factor(pais)Peru	1.666994	0.903885	1.844
## factor(pais)Philippines	4.575134	1.187240	3.854
## factor(pais)Poland	0.042422	0.717564	0.059
## factor(pais)Portugal	-1.661270	0.587557	-2.827
## factor(pais)Romania	-0.289316	0.721707	-0.401
## factor(pais)Russian Federation	2.093125	0.766138	2.732
## factor(pais)Saudi Arabia	-0.779182	0.803086	-0.970
## factor(pais)Singapore	-5.221901	1.072661	-4.868
## factor(pais)Slovakia	-2.261850	0.634800	-3.563
## factor(pais)Slovenia	-2.301782	0.596219	-3.861
## factor(pais)South Africa	2.221751	0.769873	2.886
## factor(pais)Sri Lanka	-0.144985	1.048528	-0.138
## factor(pais)Sweden	-0.943663	0.622757	-1.515
## factor(pais)Switzerland	-3.530504	0.688291	-5.129
## factor(pais)Thailand	0.811658	1.029298	0.789
## factor(pais)Tunisia	0.597933	0.930453	0.643
## factor(pais)Turkey	1.578875	0.685848	2.302
## factor(pais)Ukraine	2.761286	1.020552	2.706
## factor(pais)United Arab Emirates	-6.532789	1.582584	-4.128
## factor(pais)United Kingdom	0.193511	0.608121	0.318
## factor(pais)United States	1.388264	0.623991	2.225
## factor(pais)Uruguay	-2.378991	0.580917	-4.095
## factor(pais)Uzbekistan	2.406816	1.321156	1.822
## factor(pais)Viet Nam	1.561941	1.191821	1.311
##	Pr(> t)		
## (Intercept)	1.65e-05	***	
## XPIBc	0.062161	.	
## XINT	2.24e-06	***	
## XDEmp	0.060494	.	
## XINF	0.367257		
## XTJ_POB	< 2e-16	***	
## XPEA_POB	0.024002	*	
## XMIG_POB	0.374354		
## factor(pais)Albania	0.011445	*	
## factor(pais)Algeria	0.008267	**	
## factor(pais)Argentina	0.224357		
## factor(pais)Armenia	0.370306		
## factor(pais)Australia	0.028104	*	
## factor(pais)Austria	1.02e-05	***	
## factor(pais)Azerbaijan	0.053607	.	
## factor(pais)Bahrain	7.79e-07	***	
## factor(pais)Bangladesh	0.037485	*	
## factor(pais)Belarus	0.433986		
## factor(pais)Belgium	0.000998	***	
## factor(pais)Bolivia	0.106222		
## factor(pais)Bosnia and Herzegovina	0.507135		
## factor(pais)Brazil	0.000455	***	
## factor(pais)Bulgaria	0.261369		
## factor(pais)Canada	0.287355		
## factor(pais)Chile	0.012211	*	
## factor(pais)China	2.65e-10	***	
## factor(pais)Colombia	0.305480		

## factor(pais)Costa Rica	8.17e-07 ***
## factor(pais)Croatia	0.046131 *
## factor(pais)Czech Republic	0.011316 *
## factor(pais)Denmark	0.009753 **
## factor(pais)Dominican Republic	0.928998
## factor(pais)Ecuador	0.928580
## factor(pais)Egypt	0.002721 **
## factor(pais)Estonia	0.000125 ***
## factor(pais)Finland	0.091687 .
## factor(pais)France	0.035996 *
## factor(pais)Georgia	0.056048 .
## factor(pais)Germany	0.780665
## factor(pais)Greece	0.366834
## factor(pais)Guatemala	0.308786
## factor(pais)Hungary	0.002063 **
## factor(pais)Iceland	< 2e-16 ***
## factor(pais)India	9.63e-07 ***
## factor(pais)Indonesia	0.000531 ***
## factor(pais)Iran, Islamic Republic of	0.758462
## factor(pais)Iraq	0.000378 ***
## factor(pais)Ireland	1.67e-07 ***
## factor(pais)Israel	0.686919
## factor(pais)Italy	0.099349 .
## factor(pais)Japan	0.016201 *
## factor(pais)Jordan	1.06e-05 ***
## factor(pais)Kazakhstan	0.290403
## factor(pais)Korea, Republic of	0.346012
## factor(pais)Kuwait	1.67e-07 ***
## factor(pais)Kyrgyzstan	0.234420
## factor(pais)Lao People's Democratic Republic	0.780804
## factor(pais)Latvia	2.97e-06 ***
## factor(pais)Lebanon	0.258448
## factor(pais)Lithuania	1.26e-06 ***
## factor(pais)Luxembourg	< 2e-16 ***
## factor(pais)Macao	1.77e-07 ***
## factor(pais)Malaysia	0.879857
## factor(pais)Malta	< 2e-16 ***
## factor(pais)Mexico	0.006707 **
## factor(pais)Moldova, Republic of	0.492210
## factor(pais)Morocco	0.258960
## factor(pais)Netherlands	0.004635 **
## factor(pais>New Zealand	6.79e-07 ***
## factor(pais>Nicaragua	0.198001
## factor(pais)North Macedonia	0.133697
## factor(pais>Norway	5.13e-06 ***
## factor(pais)Pakistan	8.56e-07 ***
## factor(pais)Panama	0.000182 ***
## factor(pais)Paraguay	0.007943 **
## factor(pais)Peru	0.065965 .
## factor(pais)Philippines	0.000138 ***
## factor(pais)Poland	0.952890
## factor(pais)Portugal	0.004954 **
## factor(pais)Romania	0.688747
## factor(pais)Russian Federation	0.006604 **

```
## factor(pais)Saudi Arabia      0.332579
## factor(pais)Singapore        1.69e-06 ***
## factor(pais)Slovakia         0.000416 ***
## factor(pais)Slovenia         0.000134 ***
## factor(pais)South Africa     0.004138 **
## factor(pais)Sri Lanka        0.890100
## factor(pais)Sweden           0.130572
## factor(pais)Switzerland      4.75e-07 ***
## factor(pais)Thailand         0.430890
## factor(pais)Tunisia          0.520875
## factor(pais)Turkey           0.021899 *
## factor(pais)Ukraine          0.007140 **
## factor(pais)United Arab Emirates 4.55e-05 ***
## factor(pais)United Kingdom   0.750508
## factor(pais)United States    0.026711 *
## factor(pais)Uruguay          5.21e-05 ***
## factor(pais)Uzbekistan       0.069320 .
## factor(pais)Viet Nam         0.190843
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7425 on 361 degrees of freedom
## Multiple R-squared:  0.9254, Adjusted R-squared:  0.9052
## F-statistic: 45.7 on 98 and 361 DF, p-value: < 2.2e-16
```

Analizando significancias

```
no_significativo
```

```
## [1] "XPIBc"
## [2] "XDEmp"
## [3] "XINF"
## [4] "XMIG_POB"
## [5] "factor(pais)Argentina"
## [6] "factor(pais)Armenia"
## [7] "factor(pais)Azerbaijan"
## [8] "factor(pais)Belarus"
## [9] "factor(pais)Bolivia"
## [10] "factor(pais)Bosnia and Herzegovina"
## [11] "factor(pais)Bulgaria"
## [12] "factor(pais)Canada"
## [13] "factor(pais)Colombia"
## [14] "factor(pais)Dominican Republic"
## [15] "factor(pais)Ecuador"
## [16] "factor(pais)Finland"
## [17] "factor(pais)Georgia"
## [18] "factor(pais)Germany"
## [19] "factor(pais)Greece"
## [20] "factor(pais)Guatemala"
## [21] "factor(pais)Iran, Islamic Republic of"
## [22] "factor(pais)Israel"
## [23] "factor(pais)Italy"
```

```
## [24] "factor(pais)Kazakhstan"
## [25] "factor(pais)Korea, Republic of"
## [26] "factor(pais)Kyrgyzstan"
## [27] "factor(pais)Lao People's Democratic Republic"
## [28] "factor(pais)Lebanon"
## [29] "factor(pais)Malaysia"
## [30] "factor(pais)Moldova, Republic of"
## [31] "factor(pais)Morocco"
## [32] "factor(pais)Nicaragua"
## [33] "factor(pais)North Macedonia"
## [34] "factor(pais)Peru"
## [35] "factor(pais)Poland"
## [36] "factor(pais)Romania"
## [37] "factor(pais)Saudi Arabia"
## [38] "factor(pais)Sri Lanka"
## [39] "factor(pais)Sweden"
## [40] "factor(pais)Thailand"
## [41] "factor(pais)Tunisia"
## [42] "factor(pais)United Kingdom"
## [43] "factor(pais)Uzbekistan"
## [44] "factor(pais)Viet Nam"
```

significativo_positivos

```
## [1] "(Intercept)" "XINT"
## [3] "XTJ_POB" "XPEA_POB"
## [5] "factor(pais)Algeria" "factor(pais)Bangladesh"
## [7] "factor(pais)Brazil" "factor(pais)China"
## [9] "factor(pais)Egypt" "factor(pais)France"
## [11] "factor(pais)India" "factor(pais)Indonesia"
## [13] "factor(pais)Iraq" "factor(pais)Japan"
## [15] "factor(pais)Jordan" "factor(pais)Mexico"
## [17] "factor(pais)Pakistan" "factor(pais)Philippines"
## [19] "factor(pais)Russian Federation" "factor(pais)South Africa"
## [21] "factor(pais)Turkey" "factor(pais)Ukraine"
## [23] "factor(pais)United States"
```

significativo_negativos

```
## [1] "factor(pais)Albania" "factor(pais)Australia"
## [3] "factor(pais)Austria" "factor(pais)Bahrain"
## [5] "factor(pais)Belgium" "factor(pais)Chile"
## [7] "factor(pais)Costa Rica" "factor(pais)Croatia"
## [9] "factor(pais)Czech Republic" "factor(pais)Denmark"
## [11] "factor(pais)Estonia" "factor(pais)Hungary"
## [13] "factor(pais)Iceland" "factor(pais)Ireland"
## [15] "factor(pais)Kuwait" "factor(pais)Latvia"
## [17] "factor(pais)Lithuania" "factor(pais)Luxembourg"
## [19] "factor(pais)Macao" "factor(pais)Malta"
## [21] "factor(pais)Netherlands" "factor(pais)New Zealand"
## [23] "factor(pais)Norway" "factor(pais)Panama"
## [25] "factor(pais)Paraguay" "factor(pais)Portugal"
## [27] "factor(pais)Singapore" "factor(pais)Slovakia"
```

```
## [29] "factor(pais)Slovenia"          "factor(pais)Switzerland"
## [31] "factor(pais)United Arab Emirates" "factor(pais)Uruguay"
```

Supuestos

- normalidad

```
# summary(regresion_pais_sp)
shapiro.test(resid(regresion_pais_sp))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  resid(regresion_pais_sp)
## W = 0.96947, p-value = 3.36e-08
```

*homocedasticidad

```
bptest(regresion_pais_sp)
```

```
##
##  studentized Breusch-Pagan test
##
## data:  regresion_pais_sp
## BP = 181.3, df = 98, p-value = 6.423e-07
```

```
dwtest(regresion_pais_sp)
```

```
##
##  Durbin-Watson test
##
## data:  regresion_pais_sp
## DW = 1.7404, p-value = 6.581e-15
## alternative hypothesis: true autocorrelation is greater than 0
```

- multicolinealidad

```
vif(regresion_pais_sp)
```

```
##              GVIF Df GVIF^(1/(2*Df))
## X              23582889 7          3.362127
## factor(pais) 23582889 91          1.097763
```

By year

- El año en sí mismo no parece tener un efecto significativo en Y después de ajustar por X

```
regresion_years = lm(Y~X+factor(year))
summary(regresion_years)
```

```
##
## Call:
## lm(formula = Y ~ X + factor(year))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.7085 -1.2173 -0.0302  1.5492  5.9947
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    20.842962    2.469095   8.442 4.38e-16 ***
## XPIBc           0.013486    0.167092   0.081  0.9357
## XINT           -0.003428    0.010912  -0.314  0.7536
## XDEmp          -0.026372    0.021485  -1.227  0.2203
## XINF            0.006743    0.009473   0.712  0.4769
## XTJ_POB         0.818233    0.095910   8.531 2.25e-16 ***
## XPEA_POB        -3.968402    1.864383  -2.129  0.0338 *
## XMIG_POB       -25.760531   13.893357  -1.854  0.0644 .
## factor(year)2018  0.205572    0.299383   0.687  0.4927
## factor(year)2019  0.477842    0.306280   1.560  0.1194
## factor(year)2020  0.240108    0.316059   0.760  0.4478
## factor(year)2021  0.680640    0.320192   2.126  0.0341 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.021 on 448 degrees of freedom
## Multiple R-squared:  0.3142, Adjusted R-squared:  0.2973
## F-statistic: 18.66 on 11 and 448 DF,  p-value: < 2.2e-16
```

EDA complementario

```
# Asumiendo que df_panel tiene una columna TJ

#
# df_panel_sum <- df_panel %>%
#   group_by(year) %>%
#   summarise(TG_sum = sum(TG, na.rm = TRUE), TJ_sum = sum(TJ, na.rm = TRUE))
#
# df_panel_sum$year <- as.numeric(as.character(df_panel_sum$year))
#
# df_panel_sum$TJ_norm <- df_panel_sum$TJ_sum / max(df_panel_sum$TJ_sum) * max(df_panel_sum$TG_sum)
#
# ggplot(df_panel_sum, aes(x = year)) +
#   geom_line(aes(y = TG_sum), color = "blue") +
#   geom_point(aes(y = TG_sum), color = "blue") +
#   geom_line(aes(y = TJ_norm), color = "red") +
#   geom_point(aes(y = TJ_norm), color = "red") +
```

```
# scale_y_continuous(sec.axis = sec_axis(~ . / max(df_panel_sum$TG_sum) * max(df_panel_sum$TJ_sum),
#                                     name = " Log() Total Jugadores")) +
# labs(x = "Year", y = "Log() Total Ganancias") +
# theme(legend.position = "top") +
# theme_classic()
```

```
# Asumiendo que df_panel tiene una columna TJ
# df_sum <- df %>%
#   group_by(year) %>%
#   summarise(TG_sum = sum(TG, na.rm = TRUE), TJ_sum = sum(TJ, na.rm = TRUE))
#
# # df_sum$year <- as.numeric(as.character(df_sum$year))
#
# # Normalizar los datos de TJ para que estén en la misma escala que TG_sum
# df_sum$TJ_norm <- df_sum$TJ_sum / max(df_sum$TJ_sum) * max(df_sum$TG_sum)
#
# ggplot(df_sum, aes(x = year, y = TG_sum, group = 1)) +
#   geom_line(aes(y = TG_sum), color = "blue") +
#   geom_line(aes(y = TJ_norm), color = "red") +
#   scale_y_continuous(sec.axis = sec_axis(~ . / max(df_sum$TG_sum) * max(df_sum$TJ_sum),
#                                     name = "Total Jugadores (TJ)")) +
#   labs(x = "Year", y = "Total Ganacias (TG)") +
#   theme(legend.position = "top") +
#   ggtitle("Ingresos Totales por Año") +
#   theme_economist() +
#   scale_color_economist()
```

```
# df_panel %>%
#   group_by(pais) %>%
#   top_n(5, TG) %>%
#   arrange(desc(TG))%>%
#   distinct(pais)
```