

# Data Diagnostics

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## FULL VOTER HISTORY DIAGNOSTIC

```
vhist_full <- read_csv("full_voter_history.csv")
```

```
## Parsed with column specification:
## cols(
##   VOTER_ID = col_integer(),
##   ELECTION_TYPE = col_character(),
##   ELECTION_DATE = col_character(),
##   ELECTION_DESCRIPTION = col_character(),
##   VOTING_METHOD = col_character(),
##   PARTY = col_character(),
##   COUNTY_NAME = col_character()
## )
```

```
vhist_full$ELECTION_DATE <- mdy(vhist_full$ELECTION_DATE)
```

First I will go through all the variables. I will count unique values, count NAs, see if there is any pattern to the NAs, and clean up when necessary.

```
#Number of NAs?
```

```
sum(is.na(vhist_full$VOTER_ID))
```

```
## [1] 8
```

```
#These can probably be deleted
```

```
vhist_full <- vhist_full[!(is.na(vhist_full$VOTER_ID)), ]
```

```
#The new vhist is smaller by 8 observations!
```

```
#Number of NAs?
```

```
sum(is.na(vhist_full$ELECTION_TYPE))
```

```
## [1] 0
```

```
#No NAs!
```

```
#Checking for "weird" values
```

```
summary(as.factor(vhist_full$ELECTION_TYPE))
```

```
##      Coordinated      General      Municipal Municipal Run-off
##      8502023      21822792      659511      150317
##      Primary      Recall      School      Special
##      5673882      162447      97376      515888
##      Special District
##      61301
```

```
#All these types are straightforward, nothing to deal with!
```

```
#Number of NAs?
```

```
sum(is.na(vhist_full$ELECTION_DESCRIPTION))
```

```
## [1] 0
```

```
#No NAs!
```

```
#Number of NAs?
```

```
sum(is.na(vhist_full$ELECTION_DATE))
```

```
## [1] 0
```

```
#No NAs!
```

```
#Checking for "weird" values
```

```
unique(vhist_full$ELECTION_DATE)
```

```
## [1] "2006-11-07" "2004-11-02" "2000-11-07" "1998-11-03" "1996-11-05"
## [6] "2016-06-28" "2014-06-24" "2012-06-26" "2010-08-10" "2008-08-12"
## [11] "2004-08-10" "2002-08-13" "2015-11-03" "2013-11-05" "2009-11-03"
## [16] "2007-11-06" "2005-11-01" "2001-11-06" "1999-11-02" "1997-11-04"
## [21] "2016-11-08" "2014-11-04" "2012-11-06" "2010-11-02" "2008-11-04"
## [26] "2002-11-05" "2006-08-08" "2000-03-10" "1998-08-11" "2011-11-01"
## [31] "2003-11-04" "2000-08-08" "1994-11-08" "1992-11-03" "1996-08-13"
## [36] "1995-11-07" "1993-11-02" "1988-11-08" "1996-03-05" "1992-08-11"
## [41] "1994-08-09" "1995-06-13" "2012-04-03" "2012-08-14" "1990-11-06"
## [46] "1989-12-05" "1989-11-07" "1990-08-14" "1996-11-01" "2014-05-06"
## [51] "2013-09-10" "1988-08-09" "1991-11-05" "1990-03-20" "1998-09-10"
## [56] "1992-03-03" "1991-10-29" "1989-05-16" "1989-02-07" "1992-03-17"
## [61] "1988-12-06" "2002-11-26" "2002-08-30" "1992-11-01" "1994-08-01"
## [66] "2011-12-20" "1991-03-19" "1994-11-01" "1991-11-01" "1995-11-01"
## [71] "1996-08-01" "2016-05-03" "2010-05-04" "1995-09-01" "1995-01-01"
## [76] "2015-04-07" "2013-04-02" "2012-05-08" "1990-07-10" "1991-01-01"
## [81] "1988-05-17" "1997-11-14" "1999-05-04" "2009-08-18" "1986-11-04"
## [86] "1986-08-12" "1989-08-01" "1997-01-03" "1995-06-20" "1994-04-05"
## [91] "1998-02-23" "2001-03-06" "2015-05-05" "2011-06-07" "1900-01-01"
## [96] "1984-11-06" "1991-05-07" "1989-05-02" "2000-09-12" "1996-12-19"
## [101] "1987-05-12" "1986-04-08" "1997-04-08" "2001-11-16" "2011-05-03"
## [106] "2003-04-01" "2015-09-01" "2003-04-08" "2001-04-03" "1999-04-06"
## [111] "1996-04-02" "2004-02-24" "1992-05-19" "2003-06-03" "2003-05-06"
## [116] "1996-09-01" "1992-09-01" "1992-08-01" "1993-11-01" "2015-06-02"
## [121] "2002-11-01" "2000-11-01" "1998-11-01" "1996-08-02" "1999-11-01"
## [126] "1989-03-07" "2005-05-03" "2001-05-08" "2005-07-19" "2005-04-05"
## [131] "1997-11-01" "2003-01-01" "2001-11-01" "2006-12-12" "2011-04-05"
## [136] "2009-04-07" "2007-04-03" "1998-04-07" "2004-04-06" "2002-04-02"
## [141] "2016-04-05" "2002-05-07" "2005-07-26" "2007-01-30" "2015-03-24"
## [146] "2015-02-10" "2010-04-06" "1997-02-04" "2008-04-01" "2012-01-31"
## [151] "2014-04-01" "2011-07-12" "2016-07-26" "1991-02-05" "2007-07-10"
## [156] "2010-03-02" "1992-07-14" "1992-06-02" "1991-07-09" "2005-12-13"
## [161] "2000-05-02" "1992-09-22" "2015-03-17" "2012-01-24" "1994-09-13"
## [166] "2002-06-04" "2015-01-27" "2009-04-28" "2007-05-01" "1998-12-01"
## [171] "2010-01-12" "1994-05-02" "1996-04-09" "2001-02-06" "1998-05-05"
## [176] "2009-02-17" "2005-03-08" "2009-12-15" "2005-10-04" "2006-04-04"
## [181] "2004-01-01" "2000-06-20" "1998-01-13" "2006-05-02" "2001-02-13"
## [186] "1991-05-21" "2006-06-27" "2014-04-08" "2007-06-05" "1999-12-07"
## [191] "2009-01-20" "2016-08-09" "2009-03-03" "1996-01-01" "1989-03-21"
## [196] "2004-11-16" "2000-04-04" "1989-01-17" "2003-12-09" "2001-06-19"
## [201] "1997-11-03" "2002-08-01" "2012-05-22" "1998-08-01" "2000-08-01"
## [206] "2000-08-02" "2001-10-16" "2010-12-14" "2010-12-07" "1999-08-03"
```

```
## [211] "2008-05-06" "2006-01-01" "2000-01-01" "1996-05-07" "1999-01-01"
## [216] "1999-08-02" "2007-11-07" "1998-08-02" "2003-07-29" "1997-01-01"
## [221] "1997-08-01" "1999-08-01" "1997-08-02" "2008-12-09" "1992-07-07"
## [226] "2009-08-25" "2013-04-23" "1997-05-06" "2005-04-19" "1991-11-06"
## [231] "1996-11-07" "2003-01-14" "2007-12-18" "2008-01-29" "2003-08-02"
## [236] "1900-01-02" "1997-03-05" "1997-08-13" "1997-04-16" "2004-06-08"
## [241] "2001-06-12" "1997-01-13" "1995-08-01" "1994-09-01" "1991-09-01"
```

```
#Nothing too weird, a few elections miscoded in the 1900s
```

```
#Number of NAs?
```

```
sum(is.na(vhist_full$VOTING_METHOD))
```

```
## [1] 1513468
```

```
#A BUNCH of NAs! Let's check them out
```

```
missing_method <- vhist_full[(is.na(vhist_full$VOTING_METHOD)), ]
```

```
#Guess what county?
```

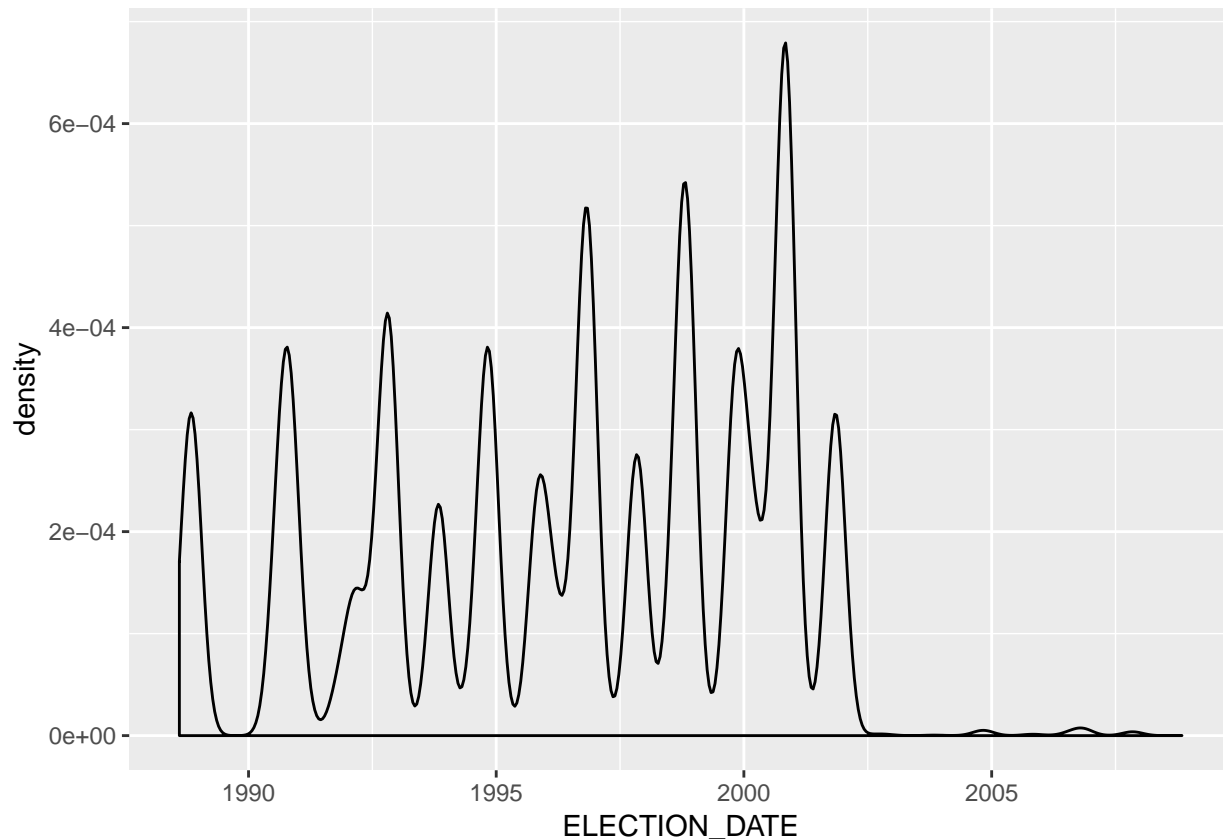
```
summary(as.factor(missing_method$COUNTY_NAME))
```

```
##      Adams      Alamosa  Archuleta    Boulder    Chaffee    Conejos
##      12         1         4         9         6         2
##      Delta      Douglas    Gilpin    Gunnison    Jackson    Jefferson
##      1         2        127        97        1136    1507147
##      Kiowa Kit Carson    La Plata      Lake Las Animas      Logan
##      453         5        113      1684        19         9
##      Mineral      Moffat  Montezuma      Morgan      Otero      Ouray
##      14         5        70         6         4         3
##      Park      Phillips    Pitkin  Rio Blanco  Rio Grande    Saguache
##      1         138         2         4         1        14
## San Miguel      Summit      Teller      Yuma
##      2367         1        10         1
```

```
#Jefferson County!
```

```
#Date distribution
```

```
ggplot(missing_method, aes(x = ELECTION_DATE)) +
  geom_density()
```



*#Note most if not all of the dates are before 2005. Therefore,  
#if I only study elections after 2012, this should not be an issue.*

*#Checking for "weird" values*

```
summary(as.factor(vhist_full$VOTING_METHOD))
```

```
##      Absentee Carry      Absentee Mail      Early Voting
##      128908             15270821          1449455
## Early Voting - DRE      In Person      In Person - DRE
##      471145             186888          156234
##      Mail Ballot      Mail Ballot - DRE      Polling Place
##      7624175           3190             10379166
##      Vote Center      Vote Center - DRE      NA's
##      261027            201060            1513468
```

*#This is very familiar...so should be fine!*

*#Number of NAs?*

```
sum(is.na(vhist_full$PARTY))
```

```
## [1] 725382
```

*#A lot of NAs!*

```
missing_party <- vhist_full[(is.na(vhist_full$PARTY)), ]
```

*#Guess what county?*

```
summary(as.factor(missing_party$COUNTY_NAME))
```

```
##      Adams      Alamosa      Arapahoe      Archuleta      Baca      Bent
```

```
##      124433      572      25076      4809      187      401
##      Boulder Broomfield Chaffee Cheyenne Clear Creek Conejos
##      26311      24092      8014      167      689      3566
##      Costilla Crowley Custer Delta Denver Dolores
##      195      278      779      1524      48414      49
##      Douglas Eagle El Paso Elbert Fremont Garfield
##      24526      3763      37207      1770      3959      2607
##      Gilpin Grand Gunnison Hinsdale Huerfano Jackson
##      2692      842      1531      39      618      244
##      Jefferson Kiowa Kit Carson La Plata Lake Larimer
##      25017      127      157      5419      2066      131793
##      Las Animas Lincoln Logan Mesa Mineral Moffat
##      1003      164      1139      53662      48      3805
##      Montezuma Montrose Morgan Otero Ouray Park
##      2052      2519      7568      5370      288      7336
##      Phillips Pitkin Prowers Pueblo Rio Blanco Rio Grande
##      112      3424      777      17465      670      596
##      Routt Saguache San Juan San Miguel Sedgwick Summit
##      4348      287      172      695      66      11098
##      Teller Washington Weld Yuma
##      9322      147      76998      318
```

```
#Actually not Jefferson...seems evenly distributed
```

```
#Date distribution
```

```
unique(missing_party$ELECTION_DATE)
```

```
## [1] "2008-11-04" "2009-08-18" "2007-11-06" "2008-08-12" "2009-01-20"
```

```
## [6] "2009-03-03" "2008-05-06" "2008-04-01" "2008-12-09"
```

```
#All elections are between 2008-2009!
```

```
#I wonder what happened...
```

```
#Checking for "weird" values generally
```

```
summary(as.factor(vhist_full$PARTY))
```

```
##      ACN AMERICAN CONSTITUTION      DEM
##      7601      6902      1231304
##      DEMOCRAT      DEMOCRATIC      GREEN
##      20646      1713032      427
##      GRN      GUN OWNERS RIGHTS      LBR
##      9284      6      32485
##      LIBERTARIAN      NO DATA      PRO-LIFE
##      3086      28985896      2
##      REFORM      REP      REPUBLICAN
##      45      1325306      2545506
##      UAF      UNAFFILIATED      UNI
##      923487      114550      590
##      NA's
##      725382
```

```
#Apart from double-coding (DEM/DEMOCRAT)
```

```
#Everything else seems fine
```

Conclusion here is that the only potentially worrying issue is some missing method values post 2012. Else the data look pretty solid.

## VOTER REG FILE DIAGNOSTICS

This has the potential to become...repetitive. I will read in all variables, describe what each one is, and then exclusively keep the useful ones.

```
reg12 <- read_csv("2012reg2.csv")

## Parsed with column specification:
## cols(
##   .default = col_character(),
##   VOTER_ID = col_integer(),
##   HOUSE_NUM = col_integer(),
##   RESIDENTIAL_ZIP_CODE = col_integer(),
##   RESIDENTIAL_ZIP_PLUS = col_integer(),
##   PRECINCT_CODE = col_double(),
##   PRECINCT_NAME = col_double(),
##   `STATE SENATE` = col_integer(),
##   `STATE HOUSE` = col_integer()
## )

## See spec(...) for full column specifications.
```

```
reg13 <- read_csv("2013reg2.csv")

## Parsed with column specification:
## cols(
##   .default = col_character(),
##   VOTER_ID = col_integer(),
##   HOUSE_NUM = col_integer(),
##   RESIDENTIAL_ZIP_CODE = col_integer(),
##   RESIDENTIAL_ZIP_PLUS = col_integer(),
##   PRECINCT_NAME = col_double(),
##   CONGRESSIONAL = col_integer(),
##   `STATE HOUSE` = col_integer()
## )

## See spec(...) for full column specifications.
```

```
reg14 <- read_csv("2014reg2.csv")

## Parsed with column specification:
## cols(
##   .default = col_character(),
##   VOTER_ID = col_integer(),
##   HOUSE_NUM = col_integer(),
##   RESIDENTIAL_ZIP_CODE = col_integer(),
##   RESIDENTIAL_ZIP_PLUS = col_integer(),
##   PRECINCT_CODE = col_double(),
##   PRECINCT_NAME = col_double(),
##   `STATE SENATE` = col_integer(),
##   `STATE HOUSE` = col_integer()
## )

## See spec(...) for full column specifications.
```

```
reg15 <- read_csv("2015reg2.csv")

## Parsed with column specification:
## cols(
```

```
## .default = col_character(),
## VOTER_ID = col_integer(),
## RESIDENTIAL_ZIP_CODE = col_integer(),
## BIRTH_YEAR = col_integer(),
## PRECINCT_CODE = col_double(),
## PRECINCT_NAME = col_double(),
## CONGRESSIONAL = col_integer(),
## `STATE HOUSE` = col_integer(),
## JUDICIAL = col_integer()
## )
## See spec(...) for full column specifications.
```

```
reg16 <- read_csv("2016reg2.csv")
```

```
## Parsed with column specification:
## cols(
## .default = col_character(),
## VOTER_ID = col_integer(),
## HOUSE_NUM = col_integer(),
## RESIDENTIAL_ZIP_CODE = col_integer(),
## PRECINCT_CODE = col_double(),
## PRECINCT_NAME = col_double(),
## `State Senate` = col_integer(),
## `State House` = col_integer()
## )
## See spec(...) for full column specifications.
```

Currently, I think I conceivably only need the following variables:

*Voter ID County Registration Date Voter Status Party Gender Birth Year Precinct Code*

The section that follow analyze these variables across the five datasets.

```
#Check for NAs
```

```
sum(is.na(reg16$VOTER_ID))
```

```
## [1] 0
```

```
sum(is.na(reg15$VOTER_ID))
```

```
## [1] 0
```

```
sum(is.na(reg14$VOTER_ID))
```

```
## [1] 0
```

```
sum(is.na(reg13$VOTER_ID))
```

```
## [1] 1
```

```
sum(is.na(reg12$VOTER_ID))
```

```
## [1] 4
```

```
#There are 5 total, 1 in 2013 and 4 in 2012.
```

```
#All in all nothing to write home about
```

```
sum(is.na(reg16$COUNTY))
```

```
## [1] 0
```

```

sum(is.na(reg15$COUNTY))

## [1] 0
sum(is.na(reg14$COUNTY))

## [1] 0
sum(is.na(reg13$COUNTY))

## [1] 4785
sum(is.na(reg12$COUNTY))

## [1] 4232
#A bit more concerning, some voters are missing county values! However,
#given the sheer number of total votes that I have at my disposal,
#8k is not a large number. I think I can disregard.

sum(is.na(reg16$REGISTRATION_DATE))

## [1] 1
sum(is.na(reg15$REGISTRATION_DATE))

## [1] 1
sum(is.na(reg14$REGISTRATION_DATE))

## [1] 3
sum(is.na(reg13$REGISTRATION_DATE))

## [1] 4789
sum(is.na(reg12$REGISTRATION_DATE))

## [1] 4238
#The numbers for 2012-13 are oddly familiar...
#Let's try pulling them out!
missing2013 <- reg13[(is.na(reg13$REGISTRATION_DATE)), ]
missing2012 <- reg12[(is.na(reg12$REGISTRATION_DATE)), ]

#NOTE THE FOLLOWING!
head(missing2012)[, 1:7]

## # A tibble: 6 x 7
##   VOTER_ID COUNTY LAST_NAME REGISTRATION_DA~ OLD_VOTER_ID HOUSE_NUM
##   <int> <chr> <chr> <chr> <chr> <int>
## 1 8207482 <NA> MCCLAVE <NA> <NA> NA
## 2 449852 <NA> KELLY <NA> <NA> NA
## 3 6305977 <NA> ROBINSON <NA> <NA> NA
## 4 1603225 <NA> WINFREE <NA> <NA> NA
## 5 8009912 <NA> DOYLE <NA> <NA> NA
## 6 6696418 <NA> DINOWITZ <NA> <NA> NA
## # ... with 1 more variable: HOUSE_SUFFIX <chr>

head(missing2013)[, 1:7]

```



```
## # A tibble: 6 x 7
##   VOTER_ID COUNTY LAST_NAME REGISTRATION_DA~ OLD_VOTER_ID HOUSE_NUM
##   <int> <chr> <chr> <chr> <chr> <int>
## 1  6.34e6 <NA> BARBER <NA> <NA> NA
## 2  1.18e5 <NA> EMANUELS~ <NA> <NA> NA
## 3  6.01e8 <NA> FORZLEY <NA> <NA> NA
## 4  4.13e6 <NA> LAUTERMI~ <NA> <NA> NA
## 5  6.36e6 <NA> BAUER <NA> <NA> NA
## 6  4.19e6 <NA> STARNES <NA> <NA> NA
## # ... with 1 more variable: HOUSE_SUFFIX <chr>
```

```
#They have NAs for almost all exept ID!
#I think this may be due to privacy concerns.
```

```
#Let's run some more diagnostics here.
#This command should count how many date entries per year
#contain more characters than necessary
sum(na.omit(!(nchar(reg12$REGISTRATION_DATE) == 10)))
```

```
## [1] 0
sum(na.omit(!(nchar(reg13$REGISTRATION_DATE) == 10)))
```

```
## [1] 0
sum(na.omit(!(nchar(reg14$REGISTRATION_DATE) == 10)))
```

```
## [1] 0
sum(na.omit(!(nchar(reg15$REGISTRATION_DATE) == 10)))
```

```
## [1] 0
sum(na.omit(!(nchar(reg16$REGISTRATION_DATE) == 10)))
```

```
## [1] 0
#All are exactly as they should be
```

```
sum(is.na(reg16$VOTER_STATUS))
```

```
## [1] 381
sum(is.na(reg15$VOTER_STATUS))
```

```
## [1] 47
sum(is.na(reg14$VOTER_STATUS))
```

```
## [1] 298
sum(is.na(reg13$VOTER_STATUS))
```

```
## [1] 5103
sum(is.na(reg12$VOTER_STATUS))
```

```
## [1] 4708
#Similarly as before, and including those hidden for privacy,
#There is a managable amount of NAs in these statuses.
```

*#Moving on to counts*

*#2012*

```
summary(as.factor(reg12$VOTER_STATUS))
```

##	Active	Inactive	UNITED STATES OF AMERICA
##	2607664	1035140	2303
##	CANADA	AUSTRALIA	UNITED KINGDOM
##	230	121	104
##	NEW ZEALAND	MEXICO	FRANCE
##	56	55	45
##	JAPAN	CHINA	SPAIN
##	45	32	31
##	ISRAEL	ITALY	GERMANY FEDERAL REPUBLIC
##	30	29	23
##	BRAZIL	BRITISH COLUMBIA CANADA	THAILAND
##	19	19	17
##	COSTA RICA	HONG KONG	IRELAND
##	16	16	15
##	ARGENTINA	CHILE	INDIA
##	13	13	12
##	SWITZERLAND	SWEDEN	UNITED ARAB EMIRATES
##	12	11	11
##	DENMARK	ENGLAND	KOREA SOUTH
##	10	10	10
##	PERU	NOVA SCOTIA CANADA	BELGIUM
##	10	9	8
##	KOREA REPUBLIC OF	ONTARIO CANADA	SCOTLAND
##	8	8	8
##	TAIWAN	AUSTRIA	INDONESIA
##	8	7	7
##	SINGAPORE	ECUADOR	MOROCCO
##	7	6	6
##	NORWAY	VIETNAM	ALBERTA CANADA
##	6	6	5
##	EGYPT	KENYA	KUWAIT
##	5	5	5
##	PHILIPPINES	RUSSIA	SAUDI ARABIA
##	5	5	5
##	BELIZE	MALAYSIA	NETHERLANDS
##	4	4	4
##	PANAMA	PUERTO RICO	QUEBEC CANADA
##	4	4	4
##	SOUTH AFRICA	WALES	BE
##	4	4	3
##	BERMUDA	GHANA	GUAM
##	3	3	3
##	POLAND	PORTUGAL	TURKEY
##	3	3	3
##	ZAMBIA	AFGHANISTAN	ANGUILLA
##	3	2	2
##	ANTARCTICA	BAHRAIN	CAMEROON
##	2	2	2
##	COLOMBIA	CZECH REPUBLIC	ESTONIA
##	2	2	2

##	FIJI	GREAT BRITAIN	GREECE
##	2	2	2
##	HONDURAS	KAZAKSTAN	MALTA
##	2	2	2
##	MOLDOVA	NEPAL	NICARAGUA
##	2	2	2
##	NIGERIA	ROMANIA	SENEGAL
##	2	2	2
##	TANZANIA	01002	02139
##	2	1	1
##	09310	09331	1510
##	1	1	1
##	1888	1890	20052
##	1	1	1
##	2794	3030	(Other)
##	1	1	41
##	NA's		
##	4708		

#2013

summary(as.factor(reg13\$VOTER\_STATUS))

##	Active	Inactive
##	2801955	752851
##	UNITED STATES OF AMERICA	CANADA
##	1582	154
##	AUSTRALIA	UNITED KINGDOM
##	92	73
##	NEW ZEALAND	MEXICO
##	43	33
##	FRANCE	JAPAN
##	31	28
##	SPAIN	CHINA
##	23	19
##	BRITISH COLUMBIA CANADA	GERMANY FEDERAL REPUBLIC
##	18	18
##	ISRAEL	ITALY
##	18	18
##	BRAZIL	THAILAND
##	16	16
##	COSTA RICA	HONG KONG
##	12	12
##	IRELAND	SWITZERLAND
##	10	10
##	ENGLAND	INDIA
##	9	9
##	NOVA SCOTIA CANADA	BELGIUM
##	9	8
##	SWEDEN	ARGENTINA
##	8	7
##	DENMARK	KOREA SOUTH
##	7	7
##	ONTARIO CANADA	INDONESIA
##	7	6
##	KOREA REPUBLIC OF	ALBERTA CANADA

##	6	5
##	CHILE	MOROCCO
##	5	5
##	PHILIPPINES	RUSSIA
##	5	5
##	SINGAPORE	TAIWAN
##	5	5
##	UNITED ARAB EMIRATES	VIETNAM
##	5	5
##	AUSTRIA	PERU
##	4	4
##	SCOTLAND	SOUTH AFRICA
##	4	4
##	AFGHANISTAN	BELIZE
##	3	3
##	ECUADOR	EGYPT
##	3	3
##	GUAM	NETHERLANDS
##	3	3
##	PUERTO RICO	WALES
##	3	3
##	BERMUDA	CZECH REPUBLIC
##	2	2
##	FIJI	GREAT BRITAIN
##	2	2
##	KENYA	MALAYSIA
##	2	2
##	NICARAGUA	NORWAY
##	2	2
##	PANAMA	POLAND
##	2	2
##	PORTUGAL	QUEBEC CANADA
##	2	2
##	Returned Mail	SAUDI ARABIA
##	2	2
##	TANZANIA	TURKEY
##	2	2
##	UGANDA	ZAMBIA
##	2	2
##	01002	02139
##	1	1
##	09310	1510
##	1	1
##	1888	1890
##	1	1
##	2794	3030
##	1	1
##	31401	31905
##	1	1
##	5000	63103
##	1	1
##	7103	80230
##	1	1
##	8803	9110

```
##          1          1
##      93410      9667
##          1          1
##      97301      AZERBAIJAN
##          1          1
##      BAHRAIN      BE
##          1          1
##      BRITISH VIRGIN ISLANDS      BULGARIA
##          1          1
##      BURMA CZECH AND SLOVAK FED REPUB
##          1          1
##      (Other)      NA's
##          21      5103
```

```
#2014
summary(as.factor(reg14$VOTER_STATUS))
```

```
##      Active      Inactive UNITED STATES OF AMERICA
##      2883194      759033      1300
##      CANADA      AUSTRALIA      UNITED KINGDOM
##      143      98      79
##      MEXICO      NEW ZEALAND      FRANCE
##      41      39      30
##      JAPAN      SPAIN      ITALY
##      24      22      21
##      BRITISH COLUMBIA CANADA      CHINA      THAILAND
##      18      18      18
##      ISRAEL      BRAZIL      COSTA RICA
##      17      15      14
##      HONG KONG GERMANY FEDERAL REPUBLIC      SWITZERLAND
##      14      12      11
##      IRELAND      NOVA SCOTIA CANADA      SINGAPORE
##      10      10      10
##      BELGIUM      ARGENTINA      ONTARIO CANADA
##      9      8      8
##      SWEDEN      DENMARK      ENGLAND
##      8      7      7
##      CHILE      INDIA      PHILIPPINES
##      6      6      6
##      SOUTH AFRICA      UNITED ARAB EMIRATES      VIETNAM
##      6      6      6
##      ALBERTA CANADA      KOREA REPUBLIC OF      KOREA SOUTH
##      5      5      5
##      MOROCCO      PERU      AUSTRIA
##      5      5      4
##      NETHERLANDS      RUSSIA      SCOTLAND
##      4      4      4
##      TAIWAN      BELIZE      EGYPT
##      4      3      3
##      GEORGIA      KENYA      MALAYSIA
##      3      3      3
##      NORWAY      AFGHANISTAN      BERMUDA
##      3      2      2
##      CROATIA      CZECH REPUBLIC      ECUADOR
##      2      2      2
```

##	FIJI	GREAT BRITAIN	GUATEMALA
##	2	2	2
##	INDONESIA	NICARAGUA	POLAND
##	2	2	2
##	PUERTO RICO	QUEBEC CANADA	Returned Mail
##	2	2	2
##	TANZANIA	TURKEY	UGANDA
##	2	2	2
##	UKRAINE	WALES	ZAMBIA
##	2	2	2
##	01002	02139	0800
##	1	1	1
##	1888	1890	20052
##	1	1	1
##	3030	31905	5000
##	1	1	1
##	5902	63103	7103
##	1	1	1
##	7562	8123	8850
##	1	1	1
##	9110	9667	98433
##	1	1	1
##	9998	ANGUILLA	BAHRAIN
##	1	1	1
##	BANGLADESH	BOLIVIA	BOTSWANA
##	1	1	1
##	BULGARIA	CAMBODIA	(Other)
##	1	1	24
##	NA's		
##	298		

#2015

```
summary(as.factor(reg15$VOTER_STATUS))
```

##	Active	AFGHANISTAN
##	2822516	1
##	ARGENTINA	AUSTRALIA
##	1	22
##	BELGIUM	BERMUDA
##	2	1
##	BRAZIL	CANADA
##	3	78
##	CHILE	CHINA
##	1	7
##	COSTA RICA	ETHIOPIA
##	2	1
##	Failed to Vote	FRANCE
##	21	2
##	GERMANY FEDERAL REPUBLIC	GREAT BRITAIN
##	5	2
##	GUAM	HONG KONG
##	1	3
##	Inactive	INDIA
##	686039	1
##	INDONESIA	IRELAND

##	1	2
##	ISRAEL	JAPAN
##	9	5
##	KOREA REPUBLIC OF	KOREA SOUTH
##	1	1
##	MEXICO MICRONESIA	FEDERATED STS
##	18	1
##	NCOA	NETHERLANDS
##	1	1
##	NEW ZEALAND	NORWAY
##	9	1
##	NOVA SCOTIA CANADA	PERU
##	1	2
##	PHILIPPINES	POLAND
##	4	2
##	REPUBLIC OF"	RUSSIA
##	3	1
##	SCOTLAND	SPAIN
##	1	3
##	SWITZERLAND	THAILAND
##	1	1
##	Undeliverable Ballot	UNITED KINGDOM
##	2	15
##	UNITED STATES OF AMERICA	VIETNAM
##	13	1
##	NA's	
##	47	

#2016

```
summary(as.factor(reg16$VOTER_STATUS))
```

##	Active	Inactive
##	3283797	550320
##	UNITED STATES OF AMERICA	CANADA
##	1243	187
##	UNITED KINGDOM	AUSTRALIA
##	100	97
##	MEXICO	SPAIN
##	45	32
##	JAPAN	NEW ZEALAND
##	31	30
##	FRANCE	GERMANY FEDERAL REPUBLIC
##	28	28
##	ISRAEL	ITALY
##	18	16
##	CHINA	HONG KONG
##	12	12
##	INDIA	IRELAND
##	12	12
##	KOREA SOUTH	NETHERLANDS
##	12	11
##	BRAZIL	THAILAND
##	9	9
##	BRITISH COLUMBIA CANADA	DENMARK
##	8	8

##	SWEDEN	CHILE
##	7	6
##	COSTA RICA	SINGAPORE
##	6	6
##	TURKEY	ENGLAND
##	6	5
##	KOREA REPUBLIC OF	NORWAY
##	5	5
##	PANAMA	PERU
##	5	5
##	PHILIPPINES	SOUTH AFRICA
##	5	5
##	TAIWAN	ARGENTINA
##	5	4
##	BELGIUM	COLOMBIA
##	4	4
##	ECUADOR	GREAT BRITAIN
##	4	4
##	GRENADA	PORTUGAL
##	4	4
##	EGYPT	GREECE
##	3	3
##	HUNGARY	POLAND
##	3	3
##	SCOTLAND	SWITZERLAND
##	3	3
##	UKRAINE	VIETNAM
##	3	3
##	AUSTRIA	BELARUS
##	2	2
##	BERMUDA	GUAM
##	2	2
##	GUATEMALA	KENYA
##	2	2
##	LUXEMBOURG	MALAWI
##	2	2
##	MOROCCO	NEPAL
##	2	2
##	PAKISTAN	QATAR
##	2	2
##	Returned Mail	TANZANIA
##	2	2
##	UGANDA	09305
##	2	1
##	20057	28310
##	1	1
##	3010	4470
##	1	1
##	49546	6100
##	1	1
##	6903	7001
##	1	1
##	7817	80401
##	1	1



```
##           80631           8428
##           1           1
##           96251          BAHRAIN
##           1           1
##           BELIZE          BENIN
##           1           1
##           BURKINA FASO      CAMBODIA
##           1           1
## CZECH AND SLOVAK FED REPUB EL SALVADOR
##           1           1
##           ETHIOPIA          GEORGIA
##           1           1
##           GUINEA            HONDURAS
##           1           1
##           INDONESIA          IRAQ
##           1           1
##           LEBANON            LITHUANIA
##           1           1
##           MALAYSIA           NICARAGUA
##           1           1
##           (Other)            NA's
##           8           381
```

*#Voter status here clearly includes absentee voters  
#and the country they live in. In fact, filtering  
#one of the years we get:*

```
country_status <- reg12[!(reg12$VOTER_STATUS %in% c("Active", "Inactive")), ]
```

*#See the problem!*

```
head(country_status)[,20:25]
```

```
## # A tibble: 6 x 6
```

```
##   VOTER_STATUS      PARTY GENDER BIRTH_YEAR PRECINCT_CODE PRECINCT_NAME
##   <chr>           <chr> <chr>   <chr>         <dbl>         <dbl>
## 1 <NA>            <NA> <NA>   <NA>           NA            NA
## 2 UNITED STATES OF AM~ <NA> UAF    Male           1993         2052619003
## 3 UNITED STATES OF AM~ <NA> REP    Female          1994         2162530070
## 4 <NA>            Acti~ <NA>   DEM            NA            1987
## 5 ITALY           <NA> DEM    Male           1961         1310816843
## 6 BELGIUM         <NA> UAF    Female          1980         1320616637
```

*#These voters' records have been a bit mishandled.  
#The variables seem "shifted" to one side  
#However, they again are very few per year,  
#So I think I can ignore them.*

```
sum(is.na(reg16$PARTY))
```

```
## [1] 1870
```

```
sum(is.na(reg15$PARTY))
```

```
## [1] 0
```

```
sum(is.na(reg14$PARTY))
```

```
## [1] 1062
```

```
sum(is.na(reg13$PARTY))
```

```
## [1] 6704
```

```
sum(is.na(reg12$PARTY))
```

```
## [1] 6740
```

```
#Similarly as before, and including those hidden for privacy,  
#There is a managable amount of NAs in these statuses.
```

```
#Moving on to counts
```

```
#2012
```

```
summary(as.factor(reg12$PARTY))
```

```
##          ACN          Active  
##          6996          262  
##          AEL          AUSTRALIA  
##          3319          2  
##          BRAZIL  BRITISH COLUMBIA  CANADA  
##          3          1  
##          CANADA          CHILE  
##          6          1  
##          CHINA          Conversion  
##          5          47  
##          COSTA RICA          DEM  
##          1          1150752  
##          EGYPT          ENGLAND  
##          1          1  
##          ETHIOPIA          Failed to Vote  
##          1          558  
##          FRANCE  GERMANY FEDERAL REPUBLIC  
##          1          1  
##          GRN          GUATEMALA  
##          10200          1  
##          HONDURAS          HONG KONG  
##          1          3  
##          Inactive          INDIA  
##          125          4  
##          IRELAND          JAPAN  
##          2          8  
##  KOREA DEMOCRATIC PEOPLES  KOREA REPUBLIC OF  
##          1          1  
##          KOREA SOUTH          LBR  
##          4          24529  
##          MEXICO  MICRONESIA  FEDERATED STS  
##          5          1  
##          NCOA          NEW ZEALAND  
##          36          1  
##          QATAR          REP  
##          1          1155877  
##          Returned Mail          SPAIN  
##          263          3  
##          TAIWAN          TUNISIA  
##          3          1
```

##	UAF	UC
##	1291100	3
##	Undeliverable Ballot	UNITED KINGDOM
##	224	3
##	UNITED STATES OF AMERICA	NA's
##	10	6740

#2013

summary(as.factor(reg13\$PARTY))

##	ACN	Active
##	7282	225
##	AUSTRALIA	BRAZIL
##	2	3
##	BRITISH COLUMBIA CANADA	CANADA
##	1	5
##	CHILE	CHINA
##	1	4
##	Conversion	COSTA RICA
##	8	2
##	DEM	EGYPT
##	1106935	1
##	ENGLAND	FRANCE
##	2	1
##	GERMANY FEDERAL REPUBLIC	GRN
##	1	9916
##	HONDURAS	HONG KONG
##	1	2
##	Inactive	JAPAN
##	33	6
##	KOREA REPUBLIC OF	KOREA SOUTH
##	1	4
##	LBR	MEXICO
##	26126	4
##	MICRONESIA FEDERATED STS	NCOA
##	1	103
##	NEW ZEALAND	REP
##	1	1120226
##	Returned Mail	SPAIN
##	99	2
##	TAIWAN	UAF
##	3	1284303
##	UC	Undeliverable Ballot
##	1	365
##	UNITED KINGDOM	UNITED STATES OF AMERICA
##	3	9
##	NA's	
##	6704	

#2014

summary(as.factor(reg14\$PARTY))

##	ACN	Active
##	9130	221
##	AUSTRALIA	BRAZIL
##	1	2

```
## BRITISH COLUMBIA CANADA CANADA
## 1 5
## CHILE CHINA
## 3 4
## COSTA RICA DEM
## 1 1117727
## EGYPT ENGLAND
## 1 1
## Failed to Vote FRANCE
## 884 2
## GERMANY FEDERAL REPUBLIC GRN
## 1 11685
## HONDURAS HONG KONG
## 1 1
## Inactive INDONESIA
## 22 1
## JAPAN KOREA REPUBLIC OF
## 2 1
## KOREA SOUTH LBR
## 4 32084
## MALAYSIA MEXICO
## 1 3
## MICRONESIA FEDERATED STS NCOA
## 1 37
## NEW ZEALAND REP
## 1 1140302
## Returned Mail SPAIN
## 45 3
## TAIWAN THAILAND
## 3 1
## UAF UKRAINE
## 1331121 1
## Undeliverable Ballot UNI
## 196 156
## UNITED KINGDOM UNITED STATES OF AMERICA
## 1 6
## NA's
## 1062
```

*#2015*

```
summary(as.factor(reg15$PARTY))
```

```
## integer(0)
```

*#THERE IS NO PARTY VARIABLE IN 2015!*

*#2016*

```
summary(as.factor(reg16$PARTY))
```

```
## ACN Active AUSTRALIA
## 11933 286 1
## BRAZIL CANADA CHINA
## 4 3 11
## COSTA RICA DEM ECUADOR
## 2 1199642 2
## ENGLAND Failed to Vote GERMANY FEDERAL REPUBLIC
```

##	1	304	1
##	GRN	HONG KONG	Inactive
##	14094	1	2
##	INDIA	INDONESIA	IRELAND
##	2	1	2
##	ITALY	JAPAN	KOREA REPUBLIC OF
##	3	1	4
##	KOREA SOUTH	LBR	MEXICO
##	1	44761	3
##	NCOA	NEW ZEALAND	REP
##	4	1	1176745
##	Returned Mail	ROMANIA	SCOTLAND
##	4	2	1
##	SOUTH AFRICA	TAJIKISTAN	THAILAND
##	1	1	4
##	UAF	UGANDA	UKRAINE
##	1385949	1	1
##	Undeliverable Ballot	UNI	UNITED ARAB EMIRATES
##	18	965	2
##	UNITED KINGDOM	UNITED STATES OF AMERICA	VENEZUELA
##	5	7	1
##	VIETNAM	ZIMBABWE	NA's
##	2	1	1870

*#Same issue as before; some files are mismatched,  
 #so values have "bled" into eachother.  
 #Need to be careful when doing any analysis that includes Party*

```
sum(is.na(reg16$GENDER))
```

```
## [1] 247
```

```
sum(is.na(reg15$GENDER))
```

```
## [1] 7
```

```
sum(is.na(reg14$GENDER))
```

```
## [1] 135
```

```
sum(is.na(reg13$GENDER))
```

```
## [1] 5048
```

```
sum(is.na(reg12$GENDER))
```

```
## [1] 4625
```

*#Similarly as before, and including those hidden for privacy,  
 #There is a managable amount of NAs in these statuses.*

*#I am not going to do counts. The results will be the same.*

The issue that arises is taht nothing seems completely clean past the first few variables, due to the issue of variable “displacement”. I will need to be very careful when using any of them for any purpose, and make sure to tidy first.