# Gerber and Green Experiment

### Gerber and Green 2003

### Exploratory data analysis

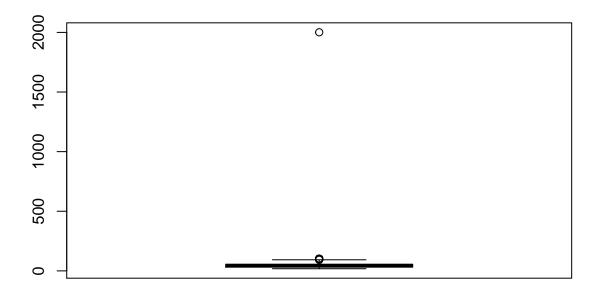
In this experiment there is total number of 18 933 individuals from 6 different cities in the United States.

#### The dependent and independent variables

The dependent variable is whether the person voted in the 6 November election in 2001. There are six independent variables; race, sex, age, party affiliation, turnout in the 2000 election and turnout in the 1999 election. The treatment indicator variable

#### Weird values

```
## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 18.00 31.00 42.00 45.25 56.00 2001.00 2487
```

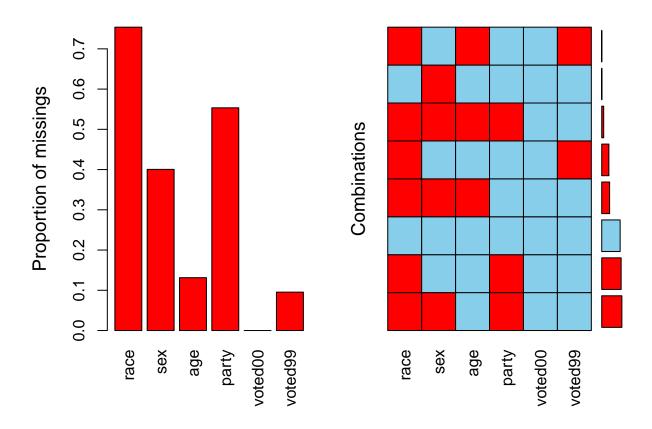


```
## city precinct zip race party sex age turf
## 13846 MINNEAPOLIS MINNEAPOLIS W- 6 P- 8 55407 2001 103
## voted01 voted00 voted99 family famsize represen reached other goaway
## 13846 1 0 0 4506 1 1 0 0 0
```

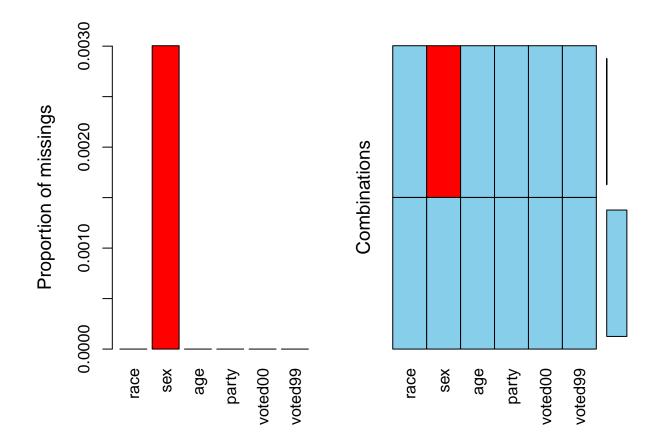
```
## nothome bad cant nothing contact treatmen primary ## 13846 0 0 0 0 0 0 0 0 0 0
```

### Missing data

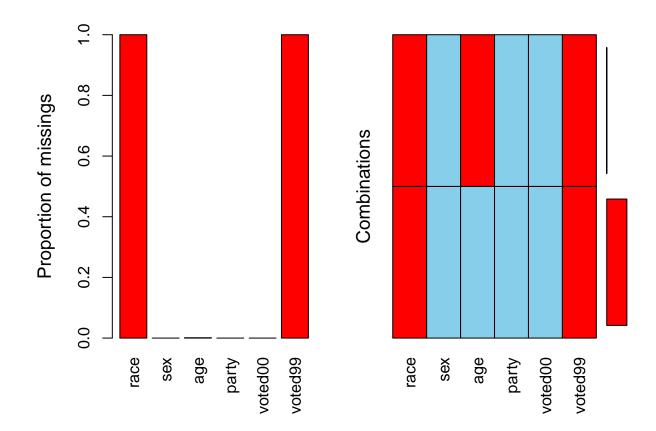
There is missing data in the experimental files - we need to recode them to the NA format to further inspect them.



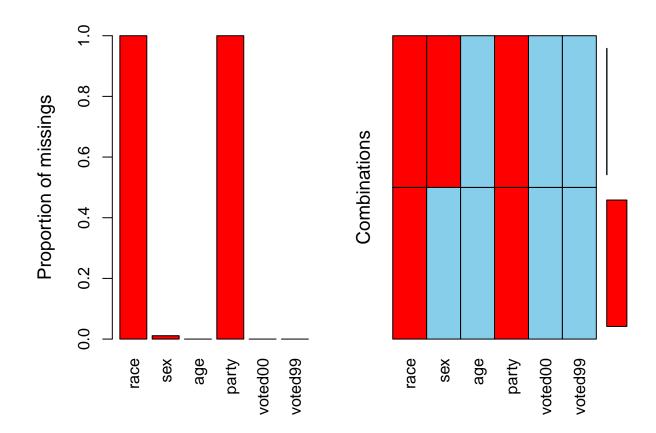
## [1] "Raleigh"



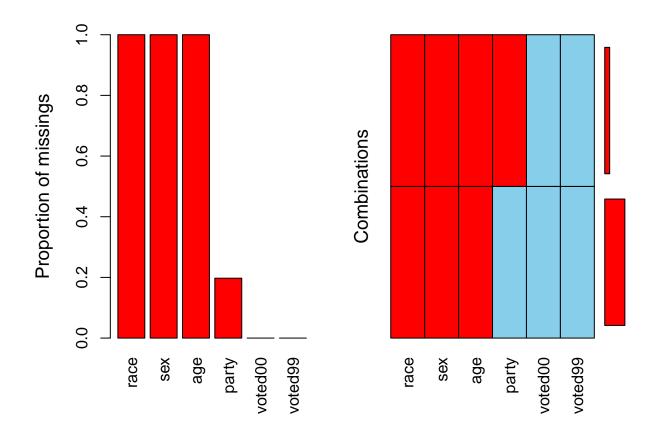
## [1] "Bridgeport"



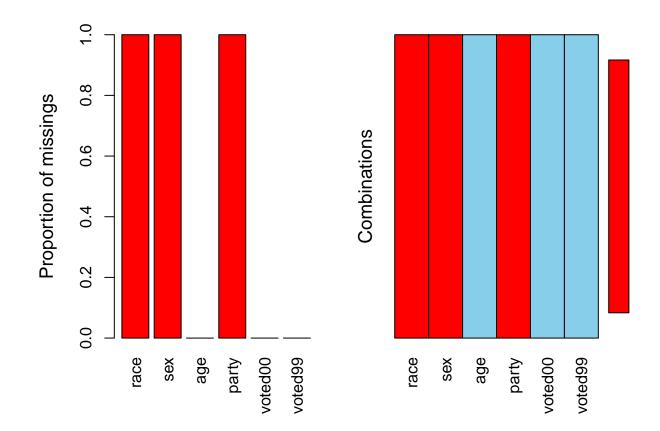
## [1] "DETROIT"



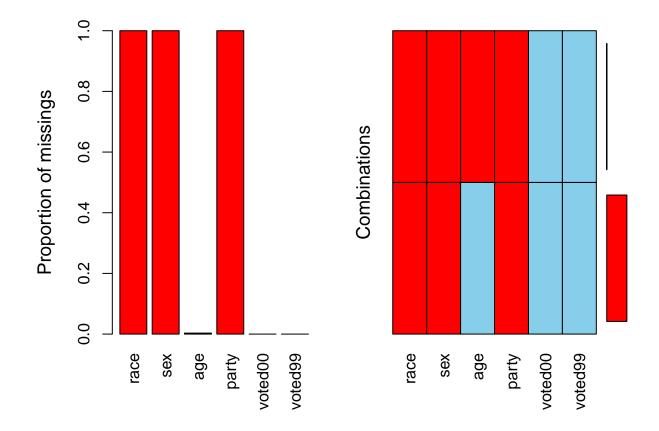
## [1] "COLUMBUS"



## [1] "MINNEAPOLIS"



## [1] "ST PAUL"



From these plots we conclude that Columbus is not useful at all. St Paul and Minneapolis only useful together. I will try this. Detroit, Raleigh, Bridgeport.

#### The rationale for putting Minneapolis and St Paul together

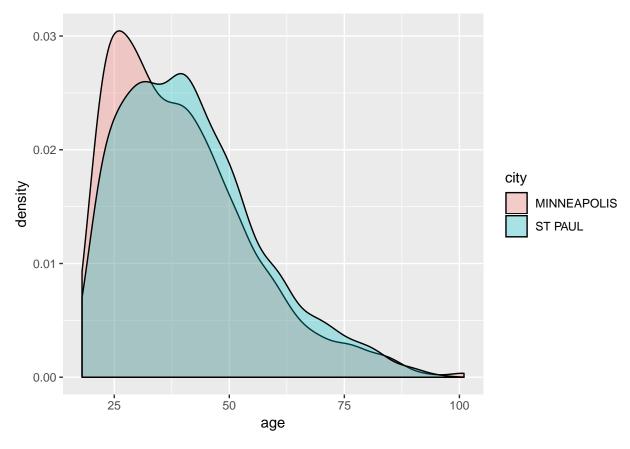
- Same state
- Same sampling composition
- x Variables to be used: age, voted00, voted99

Dealing with the one weird individual who is 2001 years old

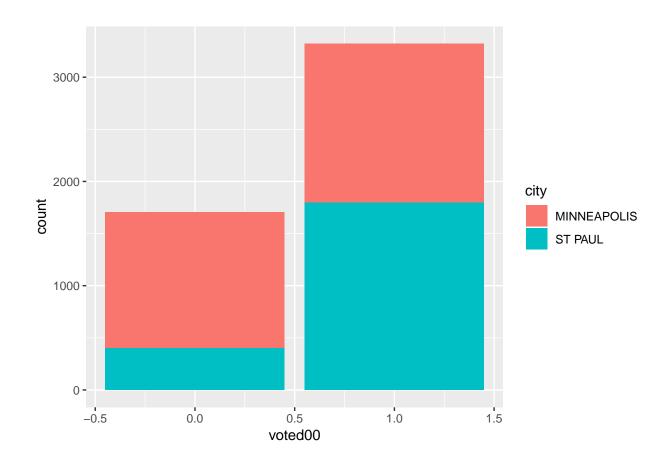
#### Detroit, Raleigh, Bridgeport

- Bridgeport seems a bit different; very low turnout. Different composition and so on.
- Detroit, Raleigh is a stretch but let's try.

## Understanding the differences in background covariates in Minnesota



## Min. 1st Qu. Mean 3rd Qu.  ${\tt Median}$ Max. 0.0000 1.0000 1.0000 0.8174 1.0000 1.0000 ## Min. 1st Qu. ## Median Mean 3rd Qu. Max. 0.0000 0.0000 1.0000 0.5389 1.0000 1.0000



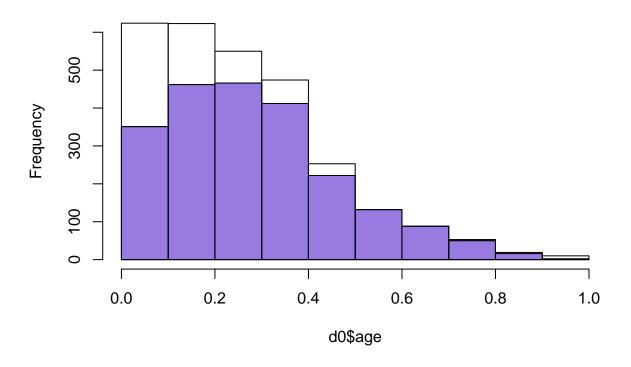
#### Rescale the variables

## Making prediction in Minnesota

First, we prepare the data. Let D=0 be Minneapolis and D=1 St Paul.

Now we run causal Match just once to see

# Histogram of d0\$age



Now we obtain the final predictions:

```
## [1] "Building trees ..."
## [1] "Tree 1"
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## [1] "Tree 99"

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## [1] "Tree 100"

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## [1] "CT"

## [1] 2201 100
```

The  $tau_1^{PRED}$  from causal match is 0.0305319 and from causal forest 0.0263373. Their respective errors are 2.008498 and 3.37339. The NPE is 6.1686766.