

BE7023 Homework 8

Mike Lape

November 21, 2018

```
#setwd("C:/Users/lapt3u/Box/UC/Fall_2018/BE7023_Adv_Biostats/adv_biostats/hw_8")
library(MatchIt)
dat <- lalonde
dim(dat)
```

```
## [1] 614 10
```

```
summary(dat)
```

```
##      treat      age      educ      black
##  Min.   :0.0000  Min.   :16.00  Min.   : 0.00  Min.   :0.0000
## 1st Qu.:0.0000  1st Qu.:20.00  1st Qu.: 9.00  1st Qu.:0.0000
##  Median :0.0000  Median :25.00  Median :11.00  Median :0.0000
##  Mean   :0.3013  Mean   :27.36  Mean   :10.27  Mean   :0.3958
## 3rd Qu.:1.0000  3rd Qu.:32.00  3rd Qu.:12.00  3rd Qu.:1.0000
##  Max.   :1.0000  Max.   :55.00  Max.   :18.00  Max.   :1.0000
##      hispan      married      nodegree      re74
##  Min.   :0.0000  Min.   :0.0000  Min.   :0.0000  Min.   : 0
## 1st Qu.:0.0000  1st Qu.:0.0000  1st Qu.:0.0000  1st Qu.: 0
##  Median :0.0000  Median :0.0000  Median :1.0000  Median :1042
##  Mean   :0.1173  Mean   :0.4153  Mean   :0.6303  Mean   :4558
## 3rd Qu.:0.0000  3rd Qu.:1.0000  3rd Qu.:1.0000  3rd Qu.:7888
##  Max.   :1.0000  Max.   :1.0000  Max.   :1.0000  Max.   :35040
##      re75      re78
##  Min.   : 0.0  Min.   : 0.0
## 1st Qu.: 0.0  1st Qu.: 238.3
##  Median :601.5  Median :4759.0
##  Mean   :2184.9  Mean   :6792.8
## 3rd Qu.:3249.0  3rd Qu.:10893.6
##  Max.   :25142.2  Max.   :60307.9
```

```
sapply(dat, class)
```

```
##      treat      age      educ      black      hispan      married      nodegree
## "integer" "integer" "integer" "integer" "integer" "integer" "integer"
##      re74      re75      re78
## "numeric" "numeric" "numeric"
```

```
head(dat)
```

```
##      treat age educ black hispan married nodegree re74 re75      re78
## NSW1     1  37  11     1      0      1         1    0    0 9930.0460
## NSW2     1  22   9     0      1      0         1    0    0 3595.8940
## NSW3     1  30  12     1      0      0         0    0    0 24909.4500
## NSW4     1  27  11     1      0      0         1    0    0 7506.1460
## NSW5     1  33   8     1      0      0         1    0    0 289.7899
## NSW6     1  22   9     1      0      0         1    0    0 4056.4940
```

```

# Optimal matching
opt <- matchit(treat ~ age + educ + black + hispan + married, data = dat,
               method = "optimal", ratio = 2)

summary(opt)

##
## Call:
## matchit(formula = treat ~ age + educ + black + hispan + married,
##         data = dat, method = "optimal", ratio = 2)
##
## Summary of balance for all data:
##           Means Treated Means Control SD Control Mean Diff eQQ Med eQQ Mean
## distance      0.5669      0.1868    0.2305    0.3801    0.4592    0.3808
## age           25.8162     28.0303   10.7867   -2.2141    1.0000    3.2649
## educ          10.3459     10.2354    2.8552    0.1105    1.0000    0.7027
## black          0.8432      0.2028    0.4026    0.6404    1.0000    0.6432
## hispan         0.0595      0.1422    0.3497   -0.0827    0.0000    0.0811
## married        0.1892      0.5128    0.5004   -0.3236    0.0000    0.3243
##           eQQ Max
## distance 0.6134
## age      10.0000
## educ      4.0000
## black      1.0000
## hispan      1.0000
## married      1.0000
##
##
## Summary of balance for matched data:
##           Means Treated Means Control SD Control Mean Diff eQQ Med eQQ Mean
## distance      0.5669      0.2113    0.2393    0.3556    0.4386    0.3568
## age           25.8162     27.6892   10.8854   -1.8730    2.0000    3.3027
## educ          10.3459     10.6568    2.7336   -0.3108    1.0000    0.6649
## black          0.8432      0.2351    0.4247    0.6081    1.0000    0.6108
## hispan         0.0595      0.1649    0.3716   -0.1054    0.0000    0.1027
## married        0.1892      0.4351    0.4964   -0.2459    0.0000    0.2432
##           eQQ Max
## distance 0.5899
## age      10.0000
## educ      4.0000
## black      1.0000
## hispan      1.0000
## married      1.0000
##
## Percent Balance Improvement:
##           Mean Diff.   eQQ Med eQQ Mean eQQ Max
## distance      6.4506     4.5021    6.2939    3.8419
## age          15.4065   -100.0000   -1.1589    0.0000
## educ        -181.2393     0.0000    5.3846    0.0000
## black         5.0493     0.0000    5.0420    0.0000
## hispan       -27.4063     0.0000  -26.6667    0.0000
## married      24.0043     0.0000   25.0000    0.0000
##
## Sample sizes:

```

```
##           Control Treated
## All           429      185
## Matched       370      185
## Unmatched      59       0
## Discarded      0       0
```

So all of our treated samples we matched, we just have 59 extra controls that were not matched.

Let's grab the match data

```
mat_dat <- match.data(opt)
```

```
dim(mat_dat)
```

```
## [1] 555 13
```

```
summary(mat_dat)
```

```
##           treat           age           educ           black
## Min.      :0.0000   Min.    :16.00   Min.    : 0.00   Min.    :0.0000
## 1st Qu.:0.0000   1st Qu.:19.00   1st Qu.: 9.00   1st Qu.:0.0000
## Median :0.0000   Median :25.00   Median :11.00   Median :0.0000
## Mean    :0.3333   Mean    :27.06   Mean    :10.55   Mean    :0.4378
## 3rd Qu.:1.0000   3rd Qu.:31.50   3rd Qu.:12.00   3rd Qu.:1.0000
## Max.    :1.0000   Max.    :55.00   Max.    :18.00   Max.    :1.0000
##           hispan        married        nodegree        re74
## Min.      :0.0000   Min.    :0.0000   Min.    :0.000   Min.    : 0.0
## 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.000   1st Qu.: 0.0
## Median :0.0000   Median :0.0000   Median :1.000   Median : 748.4
## Mean    :0.1297   Mean    :0.3532   Mean    :0.591   Mean    : 4349.2
## 3rd Qu.:0.0000   3rd Qu.:1.0000   3rd Qu.:1.000   3rd Qu.: 7389.4
## Max.    :1.0000   Max.    :1.0000   Max.    :1.000   Max.    :35040.1
##           re75           re78           distance        weights
## Min.      : 0.0   Min.    : 0.0   Min.    :0.03231   Min.    :1
## 1st Qu.: 0.0   1st Qu.: 270.5   1st Qu.:0.07828   1st Qu.:1
## Median : 469.1   Median : 4787.8   Median :0.18597   Median :1
## Mean    :2075.3   Mean    : 6822.9   Mean    :0.32982   Mean    :1
## 3rd Qu.:3074.6   3rd Qu.:10746.7   3rd Qu.:0.67423   3rd Qu.:1
## Max.    :25142.2   Max.    :60307.9   Max.    :0.75771   Max.    :1
##           subclass
## Min.      : 1
## 1st Qu.: 47
## Median : 93
## Mean    : 93
## 3rd Qu.:139
## Max.    :185
```

```
head(mat_dat)
```

```
##           treat age educ black hispan married nodegree re74 re75 re78
## NSW1         1  37  11     1      0      1      1     0   0 9930.0460
## NSW2         1  22   9     0      1      0      1     0   0 3595.8940
## NSW3         1  30  12     1      0      0      0     0   0 24909.4500
## NSW4         1  27  11     1      0      0      1     0   0 7506.1460
## NSW5         1  33   8     1      0      0      1     0   0 289.7899
## NSW6         1  22   9     1      0      0      1     0   0 4056.4940
##           distance weights subclass
```

```
## NSW1 0.4938102      1      1
## NSW2 0.1964210      1     184
## NSW3 0.7127660      1      12
## NSW4 0.6998489      1      25
## NSW5 0.6769612      1      37
## NSW6 0.6742263      1      49
```

```
table(mat_dat$subclass)
```

```
##
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 91 92 93 94 95 96 97 98 99 100 101 102 103 104 105 106 107 108
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 109 110 111 112 113 114 115 116 117 118 119 120 121 122 123 124 125 126
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 127 128 129 130 131 132 133 134 135 136 137 138 139 140 141 142 143 144
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160 161 162
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
##  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3  3
## 181 182 183 184 185
##  3  3  3  3  3
```

```
mat_dat <- mat_dat[order(mat_dat$subclass),]
```

```
head(mat_dat)
```

```
##      treat age educ black hispan married nodegree      re74      re75
## NSW1      1  37  11      1      0      1      1      0.0000      0.000
## PSID10     0  39  10      0      0      1      1 16767.4100 12022.020
## PSID115     0  20  11      0      0      1      1  5822.9410  3532.306
## NSW107     1  27  13      1      0      0      0      0.0000      0.000
## PSID90      0  29  10      0      0      1      1   713.1731  4542.048
## PSID163     0  48   8      0      0      1      1 16050.3100  2116.161
##      re78 distance weights subclass
## NSW1   9930.046 0.49381018      1      1
## PSID10  4433.180 0.03835856      1      1
## PSID115 11075.560 0.03637358      1      1
## NSW107  34099.280 0.71891686      1      2
## PSID90   7781.708 0.03643670      1      2
## PSID163 11600.150 0.03675183      1      2
```

```
mat_dat$subclass <- as.factor(mat_dat$subclass)
```

```
mat_mod <- lm(re78 ~ treat + nodegree + re74 + re75 + subclass, data = mat_dat)
summary(mat_mod)
```

```
##
## Call:
## lm(formula = re78 ~ treat + nodegree + re74 + re75 + subclass,
##     data = mat_dat)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -18416  -3756   -161    3063   35727
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.165e+03  4.211e+03   1.701  0.0897 .
## treat        8.967e+02  6.756e+02   1.327  0.1852
## nodegree    -2.109e+03  8.647e+02  -2.439  0.0152 *
## re74         3.394e-01  7.155e-02   4.744 3.01e-06 ***
## re75         1.097e-01  1.433e-01   0.765  0.4446
## subclass2    9.629e+03  5.764e+03   1.671  0.0957 .
## subclass3   -4.778e+03  5.779e+03  -0.827  0.4089
## subclass4    1.464e+03  5.819e+03   0.252  0.8015
## subclass5   -1.718e+03  5.773e+03  -0.298  0.7662
## subclass6   -4.728e+03  5.776e+03  -0.819  0.4135
## subclass7    4.055e+03  5.783e+03   0.701  0.4836
## subclass8   -1.946e+03  5.778e+03  -0.337  0.7365
## subclass9    6.295e+02  5.774e+03   0.109  0.9132
## subclass10   4.988e+03  5.778e+03   0.863  0.3885
## subclass11   2.116e+03  5.814e+03   0.364  0.7161
## subclass12   4.618e+03  5.767e+03   0.801  0.4237
## subclass13   6.097e+03  5.747e+03   1.061  0.2894
## subclass14  -5.052e+03  5.778e+03  -0.874  0.3825
## subclass15  -6.292e+03  5.756e+03  -1.093  0.2750
## subclass16  -2.122e+03  5.779e+03  -0.367  0.7137
## subclass17   6.198e+02  5.777e+03   0.107  0.9146
## subclass18  -3.477e+03  5.823e+03  -0.597  0.5508
## subclass19  -3.044e+03  5.847e+03  -0.521  0.6029
## subclass20  -1.796e+01  5.837e+03  -0.003  0.9975
## subclass21  -1.246e+03  5.773e+03  -0.216  0.8292
## subclass22  -1.226e+03  5.781e+03  -0.212  0.8321
## subclass23   5.629e+03  5.784e+03   0.973  0.3311
## subclass24  -1.024e+03  5.771e+03  -0.178  0.8592
## subclass25   1.781e+03  5.759e+03   0.309  0.7573
## subclass26  -6.864e+03  5.803e+03  -1.183  0.2376
## subclass27   1.123e+03  5.821e+03   0.193  0.8472
## subclass28   6.525e+03  5.760e+03   1.133  0.2580
## subclass29   3.813e+03  5.763e+03   0.662  0.5086
## subclass30   6.043e+02  5.773e+03   0.105  0.9167
## subclass31  -2.378e+03  5.781e+03  -0.411  0.6810
## subclass32  -4.832e+03  5.798e+03  -0.833  0.4052
## subclass33   2.488e+03  5.768e+03   0.431  0.6665
## subclass34  -3.622e+02  5.771e+03  -0.063  0.9500
## subclass35  -3.577e+03  5.764e+03  -0.621  0.5353
## subclass36  -8.932e+01  5.764e+03  -0.015  0.9876
## subclass37  -4.984e+03  5.777e+03  -0.863  0.3888
## subclass38  -1.677e+03  5.836e+03  -0.287  0.7741
## subclass39  -3.518e+03  5.777e+03  -0.609  0.5430
```

## subclass40	3.936e+03	5.802e+03	0.678	0.4980
## subclass41	-1.465e+03	5.774e+03	-0.254	0.7999
## subclass42	3.453e+02	5.795e+03	0.060	0.9525
## subclass43	8.396e+02	5.766e+03	0.146	0.8843
## subclass44	1.789e+03	5.753e+03	0.311	0.7560
## subclass45	-1.817e+03	5.773e+03	-0.315	0.7531
## subclass46	-3.698e+03	5.791e+03	-0.639	0.5235
## subclass47	2.227e+03	5.787e+03	0.385	0.7006
## subclass48	-3.725e+03	5.815e+03	-0.641	0.5221
## subclass49	2.473e+03	5.852e+03	0.423	0.6728
## subclass50	-3.501e+02	5.779e+03	-0.061	0.9517
## subclass51	-3.843e+03	5.790e+03	-0.664	0.5073
## subclass52	-5.409e+03	5.787e+03	-0.935	0.3505
## subclass53	-1.032e+01	5.765e+03	-0.002	0.9986
## subclass54	-3.574e+03	5.784e+03	-0.618	0.5371
## subclass55	-5.182e+03	5.781e+03	-0.896	0.3706
## subclass56	6.128e+01	5.777e+03	0.011	0.9915
## subclass57	-2.452e+03	5.762e+03	-0.426	0.6707
## subclass58	1.037e+03	5.762e+03	0.180	0.8572
## subclass59	9.964e+02	5.779e+03	0.172	0.8632
## subclass60	2.307e+03	5.778e+03	0.399	0.6899
## subclass61	3.317e+03	5.776e+03	0.574	0.5661
## subclass62	-2.490e+02	5.810e+03	-0.043	0.9658
## subclass63	1.689e+03	5.792e+03	0.292	0.7707
## subclass64	-1.586e+03	5.809e+03	-0.273	0.7850
## subclass65	-5.398e+03	5.776e+03	-0.934	0.3507
## subclass66	-2.368e+03	5.794e+03	-0.409	0.6829
## subclass67	2.419e+03	5.812e+03	0.416	0.6775
## subclass68	7.431e+02	5.779e+03	0.129	0.8978
## subclass69	3.385e+03	5.778e+03	0.586	0.5584
## subclass70	-2.438e+03	5.782e+03	-0.422	0.6735
## subclass71	-3.136e+03	5.772e+03	-0.543	0.5872
## subclass72	8.610e+03	5.791e+03	1.487	0.1379
## subclass73	-7.638e+02	5.805e+03	-0.132	0.8954
## subclass74	-2.110e+03	5.777e+03	-0.365	0.7151
## subclass75	1.760e+02	5.799e+03	0.030	0.9758
## subclass76	-8.214e+03	5.776e+03	-1.422	0.1559
## subclass77	-4.896e+03	5.844e+03	-0.838	0.4028
## subclass78	-7.262e+03	5.797e+03	-1.253	0.2111
## subclass79	-8.263e+03	5.764e+03	-1.434	0.1526
## subclass80	-6.157e+03	5.800e+03	-1.062	0.2891
## subclass81	-1.760e+03	5.763e+03	-0.305	0.7603
## subclass82	2.454e+03	5.812e+03	0.422	0.6731
## subclass83	5.169e+03	5.767e+03	0.896	0.3706
## subclass84	-5.915e+03	5.829e+03	-1.015	0.3109
## subclass85	-6.994e+03	5.847e+03	-1.196	0.2324
## subclass86	-3.056e+03	5.780e+03	-0.529	0.5974
## subclass87	3.275e+03	5.776e+03	0.567	0.5710
## subclass88	-6.287e+03	5.771e+03	-1.089	0.2767
## subclass89	2.138e+03	5.768e+03	0.371	0.7110
## subclass90	-1.790e+03	5.770e+03	-0.310	0.7565
## subclass91	-3.734e+03	5.777e+03	-0.646	0.5184
## subclass92	-5.305e+03	5.800e+03	-0.915	0.3609
## subclass93	7.002e+02	5.846e+03	0.120	0.9047

## subclass94	4.669e+02	5.796e+03	0.081	0.9358
## subclass95	-3.045e+03	5.786e+03	-0.526	0.5990
## subclass96	-6.264e+03	5.784e+03	-1.083	0.2795
## subclass97	-4.953e+03	5.799e+03	-0.854	0.3936
## subclass98	-8.630e+02	5.841e+03	-0.148	0.8826
## subclass99	2.509e+02	5.839e+03	0.043	0.9657
## subclass100	5.892e+03	5.791e+03	1.017	0.3096
## subclass101	-1.109e+03	5.764e+03	-0.192	0.8475
## subclass102	-1.336e+03	5.787e+03	-0.231	0.8175
## subclass103	5.930e+03	5.781e+03	1.026	0.3056
## subclass104	-4.930e+03	5.797e+03	-0.850	0.3957
## subclass105	2.022e+03	5.761e+03	0.351	0.7258
## subclass106	3.425e+03	5.764e+03	0.594	0.5527
## subclass107	8.577e+02	5.769e+03	0.149	0.8819
## subclass108	2.583e+03	5.812e+03	0.444	0.6570
## subclass109	5.060e+03	5.788e+03	0.874	0.3826
## subclass110	-2.995e+03	5.796e+03	-0.517	0.6057
## subclass111	-4.460e+03	5.782e+03	-0.771	0.4409
## subclass112	-8.593e+03	5.815e+03	-1.478	0.1403
## subclass113	-3.481e+03	5.773e+03	-0.603	0.5470
## subclass114	-7.742e+03	5.830e+03	-1.328	0.1850
## subclass115	-6.478e+03	5.811e+03	-1.115	0.2657
## subclass116	-7.127e+03	5.761e+03	-1.237	0.2168
## subclass117	1.507e+03	5.768e+03	0.261	0.7940
## subclass118	1.849e+04	5.779e+03	3.199	0.0015 **
## subclass119	-5.812e+03	5.774e+03	-1.007	0.3148
## subclass120	-2.875e+03	5.771e+03	-0.498	0.6186
## subclass121	2.827e+02	5.782e+03	0.049	0.9610
## subclass122	-4.600e+03	5.750e+03	-0.800	0.4243
## subclass123	-5.709e+02	5.767e+03	-0.099	0.9212
## subclass124	1.222e+03	5.773e+03	0.212	0.8325
## subclass125	-4.504e+03	5.786e+03	-0.778	0.4369
## subclass126	-5.240e+03	5.824e+03	-0.900	0.3689
## subclass127	5.048e+03	5.765e+03	0.876	0.3819
## subclass128	3.224e+03	5.770e+03	0.559	0.5766
## subclass129	-5.348e+03	5.763e+03	-0.928	0.3540
## subclass130	-4.894e+03	5.778e+03	-0.847	0.3975
## subclass131	1.268e+03	5.772e+03	0.220	0.8262
## subclass132	-5.194e+03	5.769e+03	-0.900	0.3685
## subclass133	-1.469e+03	5.788e+03	-0.254	0.7998
## subclass134	-3.418e+03	5.806e+03	-0.589	0.5564
## subclass135	1.136e+03	5.751e+03	0.198	0.8435
## subclass136	4.410e+03	5.764e+03	0.765	0.4447
## subclass137	-8.340e+02	5.758e+03	-0.145	0.8849
## subclass138	2.377e+03	5.754e+03	0.413	0.6798
## subclass139	-5.472e+03	5.770e+03	-0.948	0.3436
## subclass140	4.982e+03	5.762e+03	0.865	0.3878
## subclass141	-5.543e+03	5.754e+03	-0.963	0.3360
## subclass142	-2.850e+03	5.774e+03	-0.494	0.6219
## subclass143	-5.494e+03	5.815e+03	-0.945	0.3454
## subclass144	-2.523e+03	5.776e+03	-0.437	0.6626
## subclass145	-1.512e+03	5.797e+03	-0.261	0.7944
## subclass146	-6.996e+03	5.791e+03	-1.208	0.2278
## subclass147	-4.791e+02	5.812e+03	-0.082	0.9343

```

## subclass148 -5.143e+03 5.762e+03 -0.893 0.3727
## subclass149 4.578e+03 5.830e+03 0.785 0.4329
## subclass150 -1.044e+03 5.791e+03 -0.180 0.8571
## subclass151 -4.558e+02 5.788e+03 -0.079 0.9373
## subclass152 -5.012e+03 5.771e+03 -0.869 0.3857
## subclass153 -2.152e+03 5.782e+03 -0.372 0.7100
## subclass154 -8.478e+02 5.759e+03 -0.147 0.8831
## subclass155 2.005e+03 5.787e+03 0.347 0.7291
## subclass156 -6.574e+03 5.776e+03 -1.138 0.2558
## subclass157 -5.476e+03 5.767e+03 -0.950 0.3429
## subclass158 -3.840e+03 5.769e+03 -0.666 0.5061
## subclass159 2.488e+03 5.765e+03 0.432 0.6663
## subclass160 -8.627e+02 5.786e+03 -0.149 0.8816
## subclass161 -6.401e+02 5.746e+03 -0.111 0.9114
## subclass162 -1.136e+03 5.796e+03 -0.196 0.8447
## subclass163 -2.657e+03 5.845e+03 -0.455 0.6497
## subclass164 -3.285e+02 5.747e+03 -0.057 0.9544
## subclass165 -1.970e+03 5.757e+03 -0.342 0.7324
## subclass166 5.323e+03 5.753e+03 0.925 0.3554
## subclass167 -9.947e+03 5.813e+03 -1.711 0.0879
## subclass168 -6.623e+03 5.789e+03 -1.144 0.2534
## subclass169 1.868e+03 5.759e+03 0.324 0.7459
## subclass170 -8.576e+03 5.791e+03 -1.481 0.1395
## subclass171 -1.547e+03 5.771e+03 -0.268 0.7888
## subclass172 1.796e+02 5.827e+03 0.031 0.9754
## subclass173 -3.214e+03 5.760e+03 -0.558 0.5772
## subclass174 3.201e+02 5.776e+03 0.055 0.9558
## subclass175 -2.422e+03 5.775e+03 -0.419 0.6752
## subclass176 -1.957e+02 5.763e+03 -0.034 0.9729
## subclass177 -2.225e+03 5.747e+03 -0.387 0.6989
## subclass178 -3.365e+03 5.809e+03 -0.579 0.5628
## subclass179 4.838e+03 5.805e+03 0.833 0.4052
## subclass180 2.577e+02 5.747e+03 0.045 0.9643
## subclass181 -5.239e+02 5.765e+03 -0.091 0.9276
## subclass182 -9.141e+02 5.889e+03 -0.155 0.8767
## subclass183 -3.307e+03 5.828e+03 -0.568 0.5707
## subclass184 3.570e+02 5.762e+03 0.062 0.9506
## subclass185 1.993e+03 5.815e+03 0.343 0.7320
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7037 on 366 degrees of freedom
## Multiple R-squared:  0.4213, Adjusted R-squared:  0.1241
## F-statistic: 1.418 on 188 and 366 DF,  p-value: 0.002504
# R2 is low but still significant, nodegree is significant, and re74 is very significant.

# Now lets do anova
an <- anova(mat_mod, test = "chisq")
an

## Analysis of Variance Table
##
## Response: re78

```



```
##           Df      Sum Sq    Mean Sq F value    Pr(>F)
## treat      1 6.2274e+07    62273832   1.2577 0.2628171
## nodegree   1 6.5809e+08   658086062  13.2912 0.0003053 ***
## re74       1 3.6727e+09  3672707751  74.1766 < 2.2e-16 ***
## re75       1 9.6719e+07    96718720   1.9534 0.1630675
## subclass  184 8.7057e+09    47313402   0.9556 0.6331507
## Residuals 366 1.8122e+10    49513010
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# We can see that subclass is not significant and that nodegree and re74 are both significant.
```