

Untitled

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```
load("../output/feature_test.RData")
load("../output/feature_train.RData")

#the datasets after feature selection using random forest and forward
load("../output/Forward_features.RData")
load("../output/RandomForest_features.RData")
foward.index <- as.numeric(substr(forward_names,8,11))
rf.index <- as.numeric(substr(rf_names,8,11))
dat_trainf<-dat_train[foward.index]
dat_trainf$emotion_idx<-dat_train$emotion_idx
dat_trainr<-dat_train[rf.index]
dat_trainr$emotion_idx<-dat_train$emotion_idx
dat_testf<-dat_test[foward.index]
dat_testf$emotion_idx<-dat_test$emotion_idx
dat_testr<-dat_test[rf.index]
dat_testr$emotion_idx<-dat_test$emotion_idx
#the datasets after cutting off half face and using pca to reeduce dimension
load("../output/test_x.RData")
load("../output/test_y.RData")
load("../output/train_y.RData")
load("../output/train_x.RData")
test<-data.frame(test_x,emotion_idx=test_y)
train<-data.frame(train_x,emotion_idx=train_y)

# error <- as.numeric()
# for(i in 1:2){
# data.adaboost <- boosting(emotion_idx~., data=dat_train_1, mfinal=i)
# data.pred <- predict.boosting(data.adaboost,newdata = dat_test)
# error[i] <- data.pred$error
# }
# data.pred
```

Adaboost method:

```
library(adabag)

## Loading required package: rpart
## Loading required package: caret
## Loading required package: lattice
## Loading required package: ggplot2
## Loading required package: foreach
## Loading required package: doParallel
## Loading required package: iterators
```

```
## Loading required package: parallel
```

```
#adaboost method using random forest
```

```
B <- 100
```

```
data.adaboostr <- boosting(emotion_idx~., data=dat_trainr, mfinal=B)
```

```
data.predr <- predict.boosting(data.adaboostr,newdata = dat_testr)
```

```
data.predr1 <- predict.boosting(data.adaboostr,newdata = dat_trainr)
```

```
#training error
```

```
errorr1 <- data.predr1$error
```

```
errorr1
```

```
## [1] 0.8375
```

```
#testing error
```

```
errorr <- data.predr$error
```

```
errorr
```

```
## [1] 0.882
```

```
pred.indr <- as.numeric(rownames(data.pedr$confusion))
```

```
data.pedr$confusion[order(pred.indr),]
```

```
##               Observed Class
## Predicted Class  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19
##               1   2  1  5  1  3  0  3  1  0  0  0  1  2  2  3  7  1  8  0
##               2   1 11  2  1  2  1  4 14 17  4  1  2  1  0  1  1  5  0  4
##               3   3  1  2  0  0  1  0  0  0  0  1  0  0  0  0  0  2  1  0
##               4   2  1  5  9  7 14  6  2  2  1  7  4  4  1  4  1  0  0  0
##               5   0  1  1  0  3  0  3  3  0  0  1  1  0  0  0  2  1  0  3
##              11   0  0  0  0  0  0  0  0  0  0  7  0  0  0  0  0  0  0  1
##              14   5  1  7  8 12  5  7  3  7 10 10 19 12 19  9 11  7  8 18
##              17   0  1  2  0  0  0  0  2  2  0  0  0  0  0  0  1  5  1  3
##              18   3  0  1  0  2  0  0  2  0  1  1  1  1  0  0  1  2  1  2
##               Observed Class
## Predicted Class 20 21 22
##               1   0  2  0
##               2   5  4  2
##               3   2  0  0
##               4   1  1  4
##               5   0  1  0
##              11   0  0  0
##              14   8 10 14
##              17   3  3  1
##              18   3  2  1
```

```
#adaboost method using forward
```

```
B <- 100
```

```
data.adaboostf <- boosting(emotion_idx~., data=dat_trainf, mfinal=B)
```

```
data.predf <- predict.boosting(data.adaboostf,newdata = dat_testf)
```

```
data.predf1 <- predict.boosting(data.adaboostf,newdata = dat_trainf)
```

```
#training error
```

```
errorf1 <- data.predf1$error
```

```
errorf1
```

```
## [1] 0.8695
```

```
#testing error
```

```
errorf <- data.predf$error
```

```
errorf
```

```
## [1] 0.908
```

```
pred.indf<- as.numeric(rownames(data.predf$confusion))
data.predf$confusion[order(pred.indf),]
```

```
##               Observed Class
## Predicted Class  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19
##               1   3  0  2  1  0  3  2  0  0  0  0  0  1  2  0  4  0  2  2
##               2   1  7  1  1  6  2  8 13 10  1  2  4  0  0  2  4  7  1  7
##               3   4  5  5  4  3  1  5  5  8  3  2  4  3  3  4  3  6  6  4
##               4   0  0  0  1  0  2  1  0  0  0  0  0  0  0  0  1  0  0  0
##               5   1  0  0  0  2  0  0  0  0  0  0  0  0  0  0  1  1  2  0
##               8   1  2  1  0  5  0  1  5  0  0  0  0  0  0  0  1  1  2  0
##              11   0  0  0  1  0  1  1  0  0  0  6  0  0  0  0  0  0  0  0
##              14   6  3 16 11 13 12  5  4 10 12 18 20 16 17 11 10  8  6 18
##               Observed Class
## Predicted Class 20 21 22
##               1    0  0  1
##               2   10  4  2
##               3    4  3  6
##               4    0  0  0
##               5    0  0  0
##               8    2  0  0
##              11    0  0  0
##              14    6 16 13
```

```
#adaboost method using half face and pca datasets
```

```
B <- 100
```

```
data.adaboost <- boosting(emotion_idx~., data=train, mfinal=B)
```

```
data.pred <- predict.boosting(data.adaboost,newdata = test)
```

```
data.pred1<- predict.boosting(data.adaboost,newdata = train)
```

```
#training error
```

```
error1 <- data.pred1$error
```

```
error1
```

```
## [1] 0.727
```

```
#testing error
```

```
error <- data.pred$error
```

```
error
```

```
## [1] 0.76
```

```
pred.ind <- as.numeric(rownames(data.pred$confusion))
data.pred$confusion[order(pred.ind),]
```

```
##               Observed Class
## Predicted Class  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19
##               1   7  0  3  2  0  0  1  0  1  2  1  1  1  4  2  1  2  0  2
##               2   0  9  0  0  0  0  0  0  3  0  0  0  0  0  0  0  0  0  0
##               3   1  1 14  3  0  3  4  0  1  5  3  0  3  1  1  0  1  0  1
##               4   0  0  1  5  0  3  0  0  1  0  0  0  1  1  1  0  0  0  1
##               5   1  0  0  0 10  0  1  1  0  0  0  0  0  0  1  0  2  3  1
##               8   0 10  2  0  2  2  3 21  4  0  1  0  0  0  2  2  2  2  0
##               9   0  1  0  1  0  0  0  1  6  3  1  0  2  1  0  2  0  0  0
```

```
##          11  0  0  1 11  0  3  0  0  3  8  6  7  7  1  1  3  0  0  0
##          12  0  0  1  0  0  2  1  0  1  0  7  4  3  0  0  1  0  0  1
##          13  0  0  0  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0  1
##          14  1  0  3  2  4  2  9  0  0  4  1  4  1 17  9  1  2  1  7
##          16  5  3  4  5  0  4  3  1  1  0  1  1  1  0  2  8  0  0  2
##          17  5  1  1  0  8  2  5  2  2  0  0  0  0  1  4  2 13 11  2
##              Observed Class
## Predicted Class 20 21 22
##              1   2   1   3
##              2   0   0   0
##              3   0   4   4
##              4   0   0   0
##              5   1   1   0
##              8   7   1   1
##              9   0   4   0
##             11   4   7   6
##             12   1   0   0
##             13   0   0   0
##             14   4   5   3
##             16   2   1   1
##             17   2   3   1
```

svm method:

```
library(e1071)
```

```
#svm method using random forest
svm.modelr<-svm(emotion_idx~.,data=dat_trainr)
svm.predr<-predict(svm.modelr,dat_testr)
svm.predr1<-predict(svm.modelr,dat_trainr)
svm.predr
```

```
##   1   2   3   4   5   6   7   8   9  10  11  12  13  14  15  16  17  18
##   1   1   1  16   1  22  18  15   1   1   3   1  16  15   5  17   2   8
##  19  20  21  22  23  24  25  26  27  28  29  30  31  32  33  34  35  36
##   3   2   8   8   2   2   9  10   2   2   2   3   2   5   2  16   3   1
##  37  38  39  40  41  42  43  44  45  46  47  48  49  50  51  52  53  54
##  10   3  14  19   3  20   3  15  20   4  10  21  16   1  20   3  14   9
##  55  56  57  58  59  60  61  62  63  64  65  66  67  68  69  70  71  72
##   6  20  18   5   4   6   4  13   4   4   4   6  14   6  10  18   4  14
##  73  74  75  76  77  78  79  80  81  82  83  84  85  86  87  88  89  90
##   3   3   3  14  17   5  18  16  12  15   2   5   7   6   5   6   3   2
##  91  92  93  94  95  96  97  98  99 100 101 102 103 104 105 106 107 108
##   5   1  14   5   5   5  17   7   1  18  14  18  15  17  15  17   4   4
## 109 110 111 112 113 114 115 116 117 118 119 120 121 122 123 124 125 126
##   4  15   4   4   5   4   6  14   4   8  22   5  13  10   6   8  14  14
## 127 128 129 130 131 132 133 134 135 136 137 138 139 140 141 142 143 144
##  14   2   3  12   2   7  13   7   4   9   7   3   3   6   7   4  16  22
## 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160 161 162
##   3  16  15   3  15   3  17   8   2   8   8  20   8   8   8   8   8   9
## 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
##  13   9   7   3   8   2   2   8   2   8   8   8   2  21   8   9   9   3
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198
```

```
## 17 2 2 9 2 21 2 9 2 9 9 2 20 2 11 20 17 2
## 199 200 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216
## 9 22 8 21 3 20 21 13 10 10 19 13 22 10 12 13 7 15
## 217 218 219 220 221 222 223 224 225 226 227 228 229 230 231 232 233 234
## 17 21 14 13 21 1 11 12 21 11 21 11 11 11 11 11 6
## 235 236 237 238 239 240 241 242 243 244 245 246 247 248 249 250 251 252
## 6 20 20 14 14 6 4 17 2 17 22 14 4 20 1 13 11 12
## 253 254 255 256 257 258 259 260 261 262 263 264 265 266 267 268 269 270
## 13 4 13 12 1 14 14 14 14 5 19 12 15 14 12 10 13 12
## 271 272 273 274 275 276 277 278 279 280 281 282 283 284 285 286 287 288
## 11 14 14 3 14 4 9 20 4 14 1 4 14 22 14 14 22 13
## 289 290 291 292 293 294 295 296 297 298 299 300 301 302 303 304 305 306
## 14 15 13 21 14 14 20 21 14 12 14 15 14 21 12 15 13 14
## 307 308 309 310 311 312 313 314 315 316 317 318 319 320 321 322 323 324
## 4 18 4 22 1 12 1 14 14 14 14 1 14 13 4 13 1 14
## 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340 341 342
## 5 14 1 14 1 15 17 14 7 14 17 18 3 16 3 16 17 18
## 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 3 10 1 13 16 6 3 14 14 18 14 3 18 18 16 21 21 19
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378
## 17 18 8 17 17 10 21 8 8 8 17 17 3 16 14 17 20 14
## 379 380 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396
## 14 2 17 5 21 17 1 21 5 18 1 12 1 18 17 17 17 1
## 397 398 399 400 401 402 403 404 405 406 407 408 409 410 411 412 413 414
## 1 3 11 5 16 1 3 21 19 19 10 9 21 14 9 17 10 10
## 415 416 417 418 419 420 421 422 423 424 425 426 427 428 429 430 431 432
## 12 13 13 17 1 3 6 14 10 14 14 10 22 3 16 14 15 14
## 433 434 435 436 437 438 439 440 441 442 443 444 445 446 447 448 449 450
## 3 17 3 9 3 9 17 17 17 12 21 22 17 20 10 21 18 10
## 451 452 453 454 455 456 457 458 459 460 461 462 463 464 465 466 467 468
## 21 14 21 15 20 19 3 3 7 8 9 19 1 14 10 21 13 3
## 469 470 471 472 473 474 475 476 477 478 479 480 481 482 483 484 485 486
## 9 22 6 14 14 20 10 14 4 21 21 17 19 10 10 1 3 20
## 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 4 14 10 13 17 4 20 22 12 18 20 13 14 9
## Levels: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
```

```
#training accuracy
mean(svm.predr1==dat_trainr[,dim(dat_trainr)[2]])
```

```
## [1] 0.407
```

```
#testing accuracy
mean(svm.pedr==dat_testr[,dim(dat_testr)[2]])
```

```
## [1] 0.216
```

```
#svm method using forward
svm.modelf<-svm(emotion_idx~.,data=dat_trainf)
svm.predf<-predict(svm.modelf,dat_testf)
svm.predf1<-predict(svm.modelf,dat_trainf)
svm.predf
```

```
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## 1 1 18 16 5 1 17 15 18 1 3 3 14 16 5 3 2 17
## 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36
```

```

## 20  2  7  8  2  2  2  3  2  9  2  7  22  1  2  17  3  1
## 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54
## 10  3  7 20  3 17  3 16  9 16 16  3  1  1 14  3 15  8
## 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72
## 18 20  7  5  4  7  4  4  4 21  4  6 14  5 10  1 13 13
## 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90
## 14 10 17 14  3  5 18  5  1  5  2  5  7  6  5  6 17 20
## 91 92 93 94 95 96 97 98 99 100 101 102 103 104 105 106 107 108
##  5  1 17  5 15  1 17 14  1  1 14  7 14 17 14 17  4  4
## 109 110 111 112 113 114 115 116 117 118 119 120 121 122 123 124 125 126
## 13  4  4  4  6  6  6 14  4 11 22  1 14 14 14 17 14 14
## 127 128 129 130 131 132 133 134 135 136 137 138 139 140 141 142 143 144
## 14  2  3 14  3 16  7  7  4  7  7  1  2  6  7  4  1  5
## 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160 161 162
##  1 16 15  3 14 17  2  8  2  8  7  7  8  2  9  8  8  3
## 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 18  9  7 18  2  2  8 17  5 17  6  9 17 20 18  9  2 18
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198
## 17  9  2  9  2 21  9  9  2  9  9  8 20  2  5 11 17 13
## 199 200 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216
##  9 14  9 21 17 20 20 13 13 10 13 13 22 10 13 11  7  3
## 217 218 219 220 221 222 223 224 225 226 227 228 229 230 231 232 233 234
## 14 21 14 13 21 13 11 12 12 11 11 11 11 11 11 11 14
## 235 236 237 238 239 240 241 242 243 244 245 246 247 248 249 250 251 252
## 11 20 12 13 14  4 14 14  4 14 21 14 14 10  1 13 11 12
## 253 254 255 256 257 258 259 260 261 262 263 264 265 266 267 268 269 270
## 13 12 13  4  1 14 14 14 14 14 10  6 14 14  9 16 13 12
## 271 272 273 274 275 276 277 278 279 280 281 282 283 284 285 286 287 288
## 13 14 14  3 14 21  7 10 14 14  1 12 14 12 12 14 17 12
## 289 290 291 292 293 294 295 296 297 298 299 300 301 302 303 304 305 306
##  6 17  4 22 20 12 18 21 14 13 14 15 14 21 14 14  1 14
## 307 308 309 310 311 312 313 314 315 316 317 318 319 320 321 322 323 324
## 14 17 13 22  1 14  1  6  1 14 14  1 18  3  3 13  1 14
## 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340 341 342
##  5  1  1 14  1 17 17  4  4 14  3 18  3  1 18 16  2 18
## 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 18  7  1  6 16  4 16 14 14 16  1 12 18 14  4 21 17 14
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378
## 17  1 18 18 17 10 19  9  9  2 17 17  3  8 16 17 21  3
## 379 380 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396
##  4  2 20  5  4  8  1 22  5  1  1 13  1  1 17 16 18 18
## 397 398 399 400 401 402 403 404 405 406 407 408 409 410 411 412 413 414
## 18 17 11  1 13  3  3  3 19  3 10 21 21 14  2  8 10 19
## 415 416 417 418 419 420 421 422 423 424 425 426 427 428 429 430 431 432
## 14 13 20  5  1 16 14 14 10  3  5 14 17  3 10 14  3 14
## 433 434 435 436 437 438 439 440 441 442 443 444 445 446 447 448 449 450
## 20  3  3  7  3  9  8 17 17  4  9 20 17  3 19 14  1  3
## 451 452 453 454 455 456 457 458 459 460 461 462 463 464 465 466 467 468
## 21 13 21 17 20  3  3  3 13  9  8 13  1 14 10  7 14 20
## 469 470 471 472 473 474 475 476 477 478 479 480 481 482 483 484 485 486
##  2  3 14 14 14 20 10  7  4 21 21 22 19 19 10  1 10 14
## 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 16 20 10  7 12 17 12 17 12 10 10 20 21 18
## Levels: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22

```

```
#training accuracy
mean(svm.predf1==dat_trainf[,dim(dat_trainf)[2]])
```

```
## [1] 0.397
```

```
#testing accuracy
mean(svm.predf==dat_testf[,dim(dat_testf)[2]])
```

```
## [1] 0.19
```

```
#svm method using half face and pca datasets
svm.model<-svm(emotion_idx~.,data=train)
svm.pred<-predict(svm.model,test)
svm.pred1<-predict(svm.model,train)
svm.pred
```

```
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18
##  1  3  1  1  1  1  1  1  1  1  1  3  1  3  1  1  1 22
## 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36
##  1  1  2  2  2  8  2  2  2  2  2  2 20  9  9  2  8  9
## 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54
##  2  9  9  8  2  2  9  2  2  3 22 21  4  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  1  3  3 13  1  3  1  3 11 10  1 21 10
## 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90
##  3  3 13  4  4  4  4  4  4  3  4  4 10 22  6  4 13  4
## 91 92 93 94 95 96 97 98 99 100 101 102 103 104 105 106 107 108
## 13 11 12 13 11  4  3  1 10 19 13  1 13  4 11  5  5 18
## 109 110 111 112 113 114 115 116 117 118 119 120 121 122 123 124 125 126
##  5 15 18  5  5  8 15  5 15  1 20 14  1 16 14 18  5  5
## 127 128 129 130 131 132 133 134 135 136 137 138 139 140 141 142 143 144
## 15  5 14 12  3  6 12  6  6 21  6  6  6  6 11 12  6 16
## 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160 161 162
## 13 12 19 11 12 16  7  7 21  7  7  7  6  7 16  6  4  7
## 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
##  3 16  7  7 18  7 19  1 12 19 19  8 17 17 21  8  8  8
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198
##  8  8  8  8  8  8  8  8  8  8  8 17  8  8  8 18  8  8
## 199 200 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216
## 19 21  8  2  8  9 10  9  9  9  2  9  7  4  9  8  6  9
## 217 218 219 220 221 222 223 224 225 226 227 228 229 230 231 232 233 234
## 21  9  9  9  9  9  8  9  8 22 10 10  3 10 10 11 10 10
## 235 236 237 238 239 240 241 242 243 244 245 246 247 248 249 250 251 252
##  3 10  8 10 19 13 22 13  3  1 13 10 10 12 12 12 12 21
## 253 254 255 256 257 258 259 260 261 262 263 264 265 266 267 268 269 270
## 12 12 13 10 10 11 16 10  6  6 11  4  9 13 22 11 20 12
## 271 272 273 274 275 276 277 278 279 280 281 282 283 284 285 286 287 288
## 12 10 12 12 12 12 13 12 11 12 14 21  3 12 11 13  4  4
## 289 290 291 292 293 294 295 296 297 298 299 300 301 302 303 304 305 306
##  4 11 12  4  1 13  4 13 11 12 12  6 12 11 11 21 13 14
## 307 308 309 310 311 312 313 314 315 316 317 318 319 320 321 322 323 324
## 14 14 14 15 14 14 14 14 14 14 14 14 14 14 14 14 14 14
## 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340 341 342
## 14 14 20 14 14 14 19 15 14 13 15 15 14 14 21 15 15 15
## 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
```

```
## 18 13 14 13 19 17 16 14 14 1 18 18 16 16 16 16 12 8
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378
## 16 16 4 12 16 16 11 16 16 7 13 11 16 20 3 17 21 17
## 379 380 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396
## 17 18 17 7 18 19 17 8 17 14 21 19 19 13 20 18 8 16
## 397 398 399 400 401 402 403 404 405 406 407 408 409 410 411 412 413 414
## 20 18 18 7 18 18 18 7 18 17 18 16 20 1 14 14 17 22
## 415 416 417 418 419 420 421 422 423 424 425 426 427 428 429 430 431 432
## 19 19 7 19 21 21 20 19 20 19 21 7 18 1 19 19 16 18
## 433 434 435 436 437 438 439 440 441 442 443 444 445 446 447 448 449 450
## 20 9 20 21 19 21 21 20 11 21 20 20 11 7 14 21 20 16
## 451 452 453 454 455 456 457 458 459 460 461 462 463 464 465 466 467 468
## 21 22 16 17 21 21 21 21 21 22 17 21 19 19 20 7 21 20
## 469 470 471 472 473 474 475 476 477 478 479 480 481 482 483 484 485 486
## 20 7 13 21 14 19 4 13 21 19 21 19 1 19 22 11 2 22
## 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 11 22 19 19 22 11 22 10 11 11 3 10 19 3
## Levels: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
```

```
#training accuracy
mean(svm.pred1==train[,dim(train)[2]])
```

```
## [1] 0.8935
```

```
#testing accuracy
mean(svm.pred==test[,dim(test)[2]])
```

```
## [1] 0.454
```

Conclusion: For the features selected by random forest and forward methods, we can see the test accuracy is pretty low.