

Creating Model for Training Data

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```
# Libraries
library(tidyverse)
library(here)
library(cvTools)
library(coefplot)
library(glmnet)
library(GGally)
library(ROCR)

file_raw_train_data <-
  here("data/train_data_speed_dating.csv")

recode_race <-
  c(
    "1" = "Black",
    "2" = "White",
    "3" = "Hispanic",
    "4" = "API",
```

```

    "5" = "Indigenous",
    "6" = "Other"
  )

recode_goal <-
  c(
    "1" = 'Seemed like a fun night out',
    "2" = "To meet new people",
    "3" = "To get a date",
    "4" = "Looking for a serious relationship",
    "5" = "To say I did it",
    "6" = "Other"
  )

recode_date <- c(
  "1" = 'Several times a week',
  "2" = "Twice a week",
  "3" = "Once a week",
  "4" = "Twice a month",
  "5" = "Once a month",
  "6" = "Several times a year",
  "7" = "Almost never"
)

recode_go_out <-
  c(
    "1" = 'Several times a week',
    "2" = "Twice a week",
    "3" = "Once a week",
    "4" = "Twice a month",
    "5" = "Once a month",
    "6" = "Several times a year",
    "7" = "Almost never"
  )

```

Reading and cleaning Data

```

train <-
  file_raw_train_data %>%
  read_csv()

##
## -- Column specification -----
## cols(
##   .default = col_double(),
##   positin1 = col_logical(),
##   field = col_character(),

```

```
##   undergra = col_logical(),
##   mn_sat = col_logical(),
##   tuition = col_logical(),
##   from = col_character(),
##   career = col_character(),
##   attr4_1 = col_logical(),
##   sinc4_1 = col_logical(),
##   intel4_1 = col_logical(),
##   fun4_1 = col_logical(),
##   amb4_1 = col_logical(),
##   shar4_1 = col_logical(),
##   attr5_1 = col_logical(),
##   sinc5_1 = col_logical(),
##   intel5_1 = col_logical(),
##   fun5_1 = col_logical(),
##   amb5_1 = col_logical(),
##   attr1_s = col_logical(),
##   sinc1_s = col_logical()
##   # ... with 56 more columns
## )
## i Use 'spec()' for the full column specifications.
```

Added in the ratings of partner rating you and you rating the partner
diff is the avg of partner rating you - avg of you rating the partner

```
train <-
  train %>%
  mutate_at(
    vars(
      career_c,
      field_cd,
      race,
      race_o,
      zipcode,
      go_out,
      date,
      goal,
      met,
      met_o,
      samerace,
      gender,
      match
    ),
    as.character
  ) %>%
  mutate(
    field_cd =
      case_when(
        field_cd %in% c("2", "4", "5", "10") ~ "STEM",
        field_cd %in% c("1", "3", "8", "9", "11", "13") ~ "Social_Science",
        field_cd %in% c("6", "7", "14", "15", "16", "17") ~ "Humanities",
        TRUE ~ "Other"
      ),
    race = recode(race, !!! recode_race),
```

```

goal = recode(goal, !!! recode_goal),
met = if_else(met == "2", "0", met),
met_o = if_else(met_o == "2", "0", met_o),
race_o = recode(race_o, !!! recode_race),
gender = if_else(gender == "0", "female", "male"),
career_c =
  case_when(
    career_c %in% c("3", "4", "5") ~ "STEM",
    career_c %in% c("1", "7", "8", "9", "10", "11", "12", "13", "17") ~ "Social_Science",
    career_c %in% c("6") ~ "Humanities",
    TRUE ~ "Other"
  )
) %>%
mutate_at(
  vars(
    career_c,
    field_cd,
    race,
    race_o,
    goal,
    met,
    met_o,
    gender
  ),
  as.factor
)

```

Normalize the preferences, if out of 100, to percentages instead.

```

train <-
  train %>%
  rowwise() %>%
  mutate(
    sum_2_1 = sum(c(attr2_1, sinc2_1, intel2_1, fun2_1, amb2_1, shar2_1)),
    sum_4_1 = sum(c(attr4_1, sinc4_1, intel4_1, fun4_1, amb4_1, shar4_1)),
  ) %>%
  mutate_at(
    vars(attr2_1, sinc2_1, intel2_1, fun2_1, amb2_1, shar2_1),
    ~ . / sum_2_1
  ) %>%
  mutate_at(
    vars(attr4_1, sinc4_1, intel4_1, fun4_1, amb4_1, shar4_1),
    ~ . / sum_4_1
  ) %>%
  dplyr::select(-c(sum_2_1, sum_4_1))

train <-
  train %>%
  rowwise() %>%
  mutate(
    rating = sum(c(attr, sinc, intel, fun, amb, shar)),
    avg_partner_rating =

```

```

    mean(c(attr, sinc, intel, fun, amb, shar), na.rm = TRUE) / 10
  ) %>%
  mutate_at(
    vars(attr, sinc, intel, fun, amb, shar),
    ~ . / rating
  ) %>%
  select(-rating)

```

Choosing Variables

```

# Selecting only the variables that determine perception of the round of speed

# Field and career are too unevenly entered, so best to use the coded versions instead.

vars_wanted <-
  train %>%
  select(
    rowname,
    gender,
    condtn,
    match:race_o,
    like_o:age,
    field_cd,
    race:imprelig,
    goal:go_out,
    career_c:exphappy,
    dec:met,
    avg_partner_rating
  ) %>%
  select(-c(met, met_o, dec)) %>%
  summarize_all(~ sum(is.na(.))) %>%
  summarize_all(~ sum(.)) %>%
  mutate_all(~ . / 4258) %>%
  pivot_longer(
    cols = everything(),
    names_to = "var",
    values_to = "percent_missing"
  ) %>%
  filter(percent_missing < 0.5)

vars_wanted <-
  vars_wanted %>%
  arrange(desc(percent_missing)) %>%
  filter(var != "income") %>%
  pull(var)

all_training_data <-
  train %>%
  select(rowname, all_of(vars_wanted))

```

Rescaling 1-10 scale to 0-1 scale

```
all_training_data <-
  all_training_data %>%
  mutate_at(
    vars(prob, prob_o, like, like_o, imprace, imprelig, sports:exphappy),
    ~ . / 10
  ) %>%
  mutate(
    match = as.double(match)
  )

cor_data <-
  all_training_data %>%
  drop_na()
```

Checking for correlations

```
# Convert data to numeric
corr <-
  cor_data %>%
  mutate_all(as.integer)

# ggcorr(corr,
#   method = c("pairwise", "spearman"),
#   nbreaks = 6,
#   hjust = 0.8,
#   label = TRUE,
#   label_size = 3,
#   color = "grey50")
```

Create a train and a validate set

```
random_rows <-
  all_training_data %>%
  pull(rowname) %>%
  as_tibble %>%
  sample_frac(0.8) %>%
  mutate(
    value = as.character(value)
  ) %>%
  pull(value)

train_model_data <-
  all_training_data %>%
  as_tibble() %>%
  filter(rowname %in% random_rows)
```

```

test_model_data <-
  all_training_data %>%
  anti_join(train_model_data, by = "rowname") %>%
  select(-rowname)

lasso_model_data <-
  train_model_data %>%
  select(-rowname) %>%
  mutate_at(vars(condtn), as.factor) %>%
  drop_na()

```

Using lasso to weed out extraneous vars

```

# convert the data to a design matrix
X = model.matrix(match ~ 0 + ., lasso_model_data)
Y = lasso_model_data %>% pull(match)

# standardize
X = scale(X)
Y = scale(Y)

set.seed(1244)

# create training and test set
train.ind = sample(nrow(X), round(nrow(X)/2))
X.train = X[train.ind,]
X.test = X[-train.ind,]
Y.train = Y[train.ind]
Y.test = Y[-train.ind]

# set lambda sequence to use for lasso and ridge
lambdas = 10^seq(-2, 1.5, 0.1)

#lasso
fm.lasso = glmnet(X.train, Y.train, alpha = 1, lambda = lambdas, thresh = 1e-12)

fm.lasso %>%
  summary()

```

```

##           Length Class      Mode
## a0           36   -none-   numeric
## beta        2484 dgCMatrix S4
## df           36   -none-   numeric
## dim           2   -none-   numeric
## lambda        36   -none-   numeric
## dev.ratio     36   -none-   numeric
## nulldev        1   -none-   numeric
## npasses        1   -none-   numeric
## jerr           1   -none-   numeric

```

```
## offset      1  -none-    logical
## call        6  -none-    call
## nobs        1  -none-    numeric
```

```
cv <- cv.glmnet(X.train, Y.train)
```

```
cv$lambda.min
```

```
## [1] 0.02421449
```

```
covars_wanted <-
  extract.coef(cv) %>%
  select(coef = Coefficient) %>%
  filter(coef != "(Intercept)") %>%
  pull(coef)
```

```
covars_wanted <-
  covars_wanted %>%
  as_tibble() %>%
  filter(value %in% colnames(train)) %>%
  pull(value)
```

```
covars_wanted
```

```
## [1] "attr"          "sinc"          "prob"
## [4] "prob_o"        "like"          "like_o"
## [7] "avg_partner_rating" "imprace"      "tvsports"
## [10] "art"           "gaming"        "clubbing"
## [13] "tv"            "shopping"      "age"
```

```
# train_model_data <-
#   train_model_data %>%
#   select(match, all_of(covars_wanted))
```

Data used

```
# Add in field_cd and career_c b/c rep as factors in covars_wanted
train_model_data <-
  train_model_data %>%
  select(match, attr, sinc, fun, prob, prob_o, like, like_o, avg_partner_rating, imprace, tvsports, clubbing)
```

Functions

Accuracy

```
accuracy <- function(table_mat) {
  sum(diag(table_mat)) / sum(table_mat)
}
```


Precision and

```
precision <- function(matrix) {  
  tp <- matrix[2, 2]  
  fp <- matrix[1, 2]  
  return(tp / (tp + fp))  
}  
  
recall <- function(matrix) {  
  tp <- matrix[2, 2]  
  fn <- matrix[2, 1]  
  return(tp / (tp + fn))  
}
```

Linear Regression

With all lasso covariates

```
m1 <-  
  glm(  
    match ~ .,  
    data = train_model_data,  
    family = 'binomial'  
  )  
  
predict <- predict(m1, test_model_data, type = 'response')  
# confusion matrix  
table_mat <- table(test_model_data$match, predict > 0.5)  
table_mat
```

```
##  
##      FALSE TRUE  
##  0    551   13  
##  1     93   33
```

```
accuracy(table_mat)
```

```
## [1] 0.8463768
```

```
precision(table_mat)
```

```
## [1] 0.7173913
```

```
recall(table_mat)
```

```
## [1] 0.2619048
```

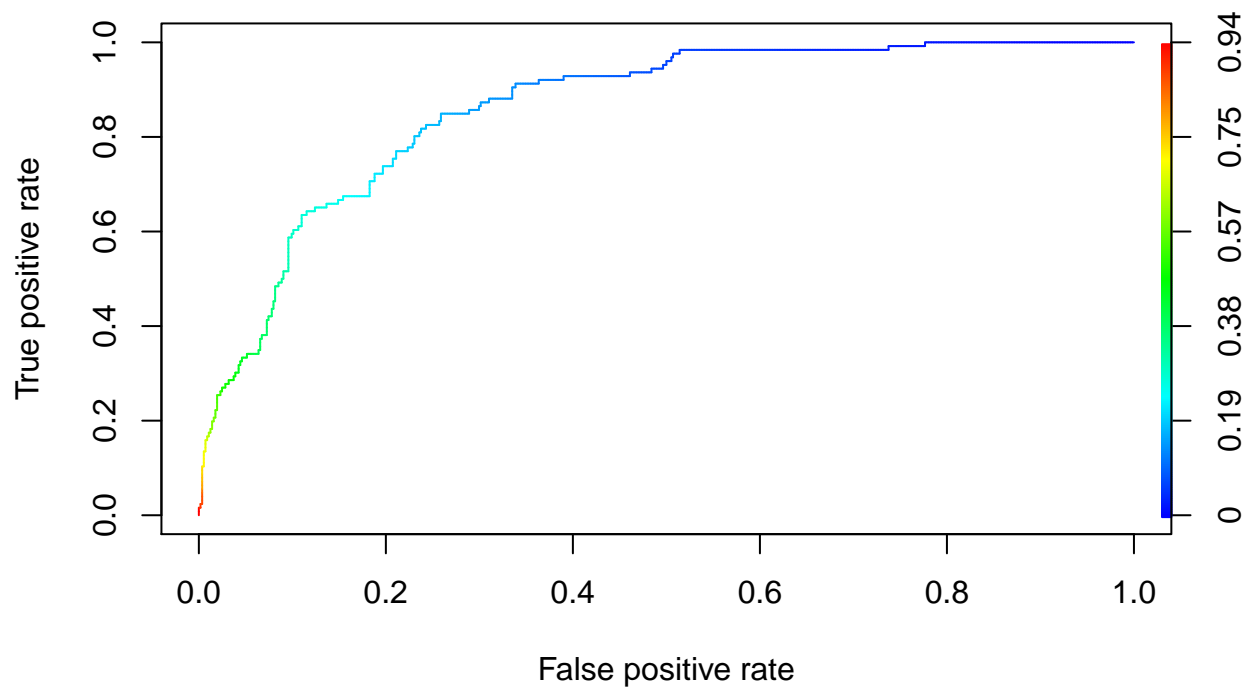
ROC curve

```
index <-  
  predict[!is.na(predict)] %>%  
  as.data.frame() %>%  
  add_rownames() %>%  
  pull(rowname)
```

```
## Warning: 'add_rownames()' is deprecated as of dplyr 1.0.0.  
## Please use 'tibble::rownames_to_column()' instead.  
## This warning is displayed once every 8 hours.  
## Call 'lifecycle::last_warnings()' to see where this warning was generated.
```

```
t <-  
  test_model_data %>%  
  add_rownames() %>%  
  filter(rowname %in% index)
```

```
ROCRpred <- prediction(predict[!is.na(predict)], t$match)  
ROCRperf <- performance(ROCRpred, 'tpr', 'fpr')  
plot(ROCRperf, colorize = TRUE, text.adj = c(-0.2, 1.7))
```



Creating a model with interaction terms

```
m2 <-
  glm(
    match ~ . + like:avg_partner_rating + attr:like_o,
    data = train_model_data,
    family = 'binomial'
  )
```

```
m2 %>%
  summary()
```

```
##
## Call:
## glm(formula = match ~ . + like:avg_partner_rating + attr:like_o,
##      family = "binomial", data = train_model_data)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.0245  -0.5810  -0.2958  -0.0780   3.3629
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -15.09438     2.40489  -6.277 3.46e-10 ***
```

```
## attr          -16.43536      8.36017   -1.966   0.04931 *
## sinc          -3.05433      2.44250   -1.250   0.21112
## fun           4.53784      2.53057    1.793   0.07294 .
## prob          1.65875      0.35865    4.625  3.75e-06 ***
## prob_o        1.00770      0.32328    3.117   0.00183 **
## like          14.29050      2.46943    5.787  7.17e-09 ***
## like_o        -0.55933      1.94250   -0.288   0.77339
## avg_partner_rating 13.15929      2.59113    5.079  3.80e-07 ***
## imprace       -0.58554      0.22462   -2.607   0.00914 **
## tvsports      -0.58610      0.23247   -2.521   0.01170 *
## clubbing       0.54193      0.25425    2.131   0.03305 *
## movies        -0.12667      0.37313   -0.339   0.73425
## shopping      -0.32508      0.24782   -1.312   0.18959
## age_o         0.01314      0.01820    0.722   0.47047
## like:avg_partner_rating -15.52744      3.33878   -4.651  3.31e-06 ***
## attr:like_o    33.20009     11.73776    2.828   0.00468 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 2495.3  on 2721  degrees of freedom
## Residual deviance: 1837.3  on 2705  degrees of freedom
##    (684 observations deleted due to missingness)
## AIC: 1871.3
##
## Number of Fisher Scoring iterations: 7
```

```
predict <- predict(m2, test_model_data, type = 'response')
# confusion matrix
table_mat <- table(test_model_data$match, predict > 0.5)
table_mat
```

```
##
##      FALSE TRUE
##  0    549   15
##  1     95   31
```

```
accuracy(table_mat)
```

```
## [1] 0.8405797
```

```
precision(table_mat)
```

```
## [1] 0.673913
```

```
recall(table_mat)
```

```
## [1] 0.2460317
```

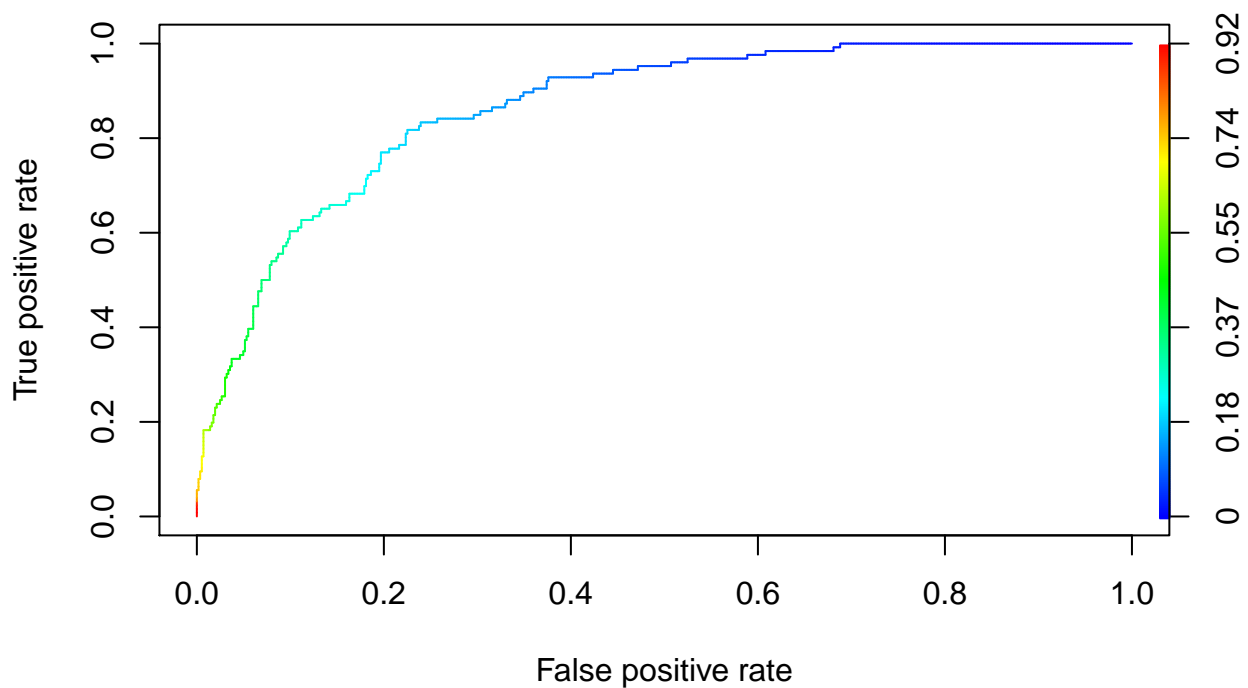
```

index <-
  predict[!is.na(predict)] %>%
  as.data.frame() %>%
  add_rownames() %>%
  pull(rowname)

t <-
  test_model_data %>%
  add_rownames() %>%
  filter(rowname %in% index)

ROCRpred <- prediction(predict[!is.na(predict)], t$match)
ROCRperf <- performance(ROCRpred, 'tpr', 'fpr')
plot(ROCRperf, colorize = TRUE, text.adj = c(-0.2, 1.7))

```



```

m3 <-
  glm(
    match ~ . + .:. ,
    data = train_model_data,
    family = 'binomial'
  )

summary(m3)

```

```
##
## Call:
## glm(formula = match ~ . + .:., family = "binomial", data = train_model_data)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.3465  -0.5523  -0.2558  -0.0569   3.2706
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      2.562844   13.054234    0.196  0.84436
## attr           -56.841880   31.455184   -1.807  0.07075 .
## sinc           -24.747707   34.378208   -0.720  0.47161
## fun            -30.784516   39.747810   -0.774  0.43864
## prob           10.019098    6.953780    1.441  0.14964
## prob_o          9.082191    6.335384    1.434  0.15170
## like           13.540555   11.128836    1.217  0.22371
## like_o          -7.564276    8.549030   -0.885  0.37626
## avg_partner_rating -15.269994   14.293980   -1.068  0.28539
## imprace         9.519474    4.561240    2.087  0.03688 *
## tvsports        -4.044074    4.913000   -0.823  0.41043
## clubbing         0.050032    5.245624    0.010  0.99239
## movies          -1.240649    7.671741   -0.162  0.87153
## shopping         -3.726093    5.303631   -0.703  0.48233
## age_o           -0.146225    0.349461   -0.418  0.67563
## attr:sinc       126.684808   63.019932    2.010  0.04441 *
## attr:fun        -37.906535   68.661470   -0.552  0.58089
## attr:prob        -2.546082   12.293240   -0.207  0.83592
## attr:prob_o      -7.663772   12.383565   -0.619  0.53600
## attr:like         3.125321   19.002320    0.164  0.86936
## attr:like_o      49.942301   16.196762    3.083  0.00205 **
## attr:avg_partner_rating 25.525084   25.618929    0.996  0.31909
## attr:imprace     -6.013046    8.549871   -0.703  0.48187
## attr:tvsports    12.997879    8.883784    1.463  0.14344
## attr:clubbing    -15.675446    9.684145   -1.619  0.10552
## attr:movies      -17.965598   12.962466   -1.386  0.16576
## attr:shopping     -0.002026    9.286829    0.000  0.99983
## attr:age_o        0.767650    0.713843    1.075  0.28221
## sinc:fun         -7.404374   73.300368   -0.101  0.91954
## sinc:prob        -2.939977   14.811377   -0.198  0.84266
## sinc:prob_o      -17.444293   14.110920   -1.236  0.21637
## sinc:like         5.214982   23.900245    0.218  0.82727
## sinc:like_o      24.004323   18.794308    1.277  0.20153
## sinc:avg_partner_rating 12.593972   29.897225    0.421  0.67358
## sinc:imprace     -11.089449    9.755364   -1.137  0.25564
## sinc:tvsports     15.017989   10.362597    1.449  0.14727
## sinc:clubbing    -14.798400   10.817342   -1.368  0.17130
## sinc:movies       12.073123   15.991392    0.755  0.45026
## sinc:shopping      5.786778   10.652504    0.543  0.58697
## sinc:age_o        -0.760253    0.795965   -0.955  0.33951
## fun:prob         -15.882649   15.202179   -1.045  0.29613
## fun:prob_o       -0.209788   14.549697   -0.014  0.98850
## fun:like         -12.151115   21.925781   -0.554  0.57945
## fun:like_o        3.789919   21.031678    0.180  0.85700
```

## fun:avg_partner_rating	60.775067	30.163908	2.015	0.04392	*
## fun:imprace	-7.899051	10.079435	-0.784	0.43323	
## fun:tvsports	-5.880713	9.642161	-0.610	0.54193	
## fun:clubbing	22.095171	10.854835	2.036	0.04180	*
## fun:movies	-19.475894	16.835081	-1.157	0.24733	
## fun:shopping	-8.832325	11.132946	-0.793	0.42757	
## fun:age_o	1.203518	0.926750	1.299	0.19407	
## prob:prob_o	0.156190	1.814301	0.086	0.93140	
## prob:like	-1.286061	3.282279	-0.392	0.69519	
## prob:like_o	-4.211013	2.785797	-1.512	0.13064	
## prob:avg_partner_rating	-1.042306	4.190312	-0.249	0.80356	
## prob:imprace	-0.166862	1.460711	-0.114	0.90905	
## prob:tvsports	-3.642695	1.465607	-2.485	0.01294	*
## prob:clubbing	1.328363	1.649540	0.805	0.42065	
## prob:movies	1.979507	2.366999	0.836	0.40299	
## prob:shopping	-1.580659	1.596352	-0.990	0.32209	
## prob:age_o	0.003581	0.122297	0.029	0.97664	
## prob_o:like	-4.056636	4.037084	-1.005	0.31497	
## prob_o:like_o	-4.185983	2.038347	-2.054	0.04001	*
## prob_o:avg_partner_rating	2.466120	4.779503	0.516	0.60587	
## prob_o:imprace	-0.793209	1.334133	-0.595	0.55214	
## prob_o:tvsports	1.192584	1.407076	0.848	0.39668	
## prob_o:clubbing	-0.863729	1.544249	-0.559	0.57594	
## prob_o:movies	-2.146037	2.357755	-0.910	0.36272	
## prob_o:shopping	1.183620	1.449374	0.817	0.41413	
## prob_o:age_o	0.066822	0.101833	0.656	0.51170	
## like:like_o	8.959405	5.191781	1.726	0.08440	.
## like:avg_partner_rating	-18.116225	4.351411	-4.163	3.14e-05	***
## like:imprace	0.747291	2.425024	0.308	0.75796	
## like:tvsports	2.889923	2.632532	1.098	0.27230	
## like:clubbing	-5.878522	2.928154	-2.008	0.04469	*
## like:movies	0.557524	4.034072	0.138	0.89008	
## like:shopping	-0.325484	2.858456	-0.114	0.90934	
## like:age_o	0.061479	0.216330	0.284	0.77627	
## like_o:avg_partner_rating	2.282481	6.388288	0.357	0.72087	
## like_o:imprace	-1.397034	1.816362	-0.769	0.44181	
## like_o:tvsports	-1.215176	1.893804	-0.642	0.52109	
## like_o:clubbing	1.418560	2.114694	0.671	0.50234	
## like_o:movies	1.130330	3.174828	0.356	0.72182	
## like_o:shopping	1.034812	2.028433	0.510	0.60994	
## like_o:age_o	-0.179254	0.132567	-1.352	0.17632	
## avg_partner_rating:imprace	1.081051	3.059304	0.353	0.72381	
## avg_partner_rating:tvsports	2.206865	3.211487	0.687	0.49197	
## avg_partner_rating:clubbing	2.389709	3.636994	0.657	0.51114	
## avg_partner_rating:movies	7.133484	5.032193	1.418	0.15632	
## avg_partner_rating:shopping	6.231323	3.526703	1.767	0.07725	.
## avg_partner_rating:age_o	0.018684	0.272873	0.068	0.94541	
## imprace:tvsports	-0.813929	1.021921	-0.796	0.42576	
## imprace:clubbing	0.417600	1.036140	0.403	0.68692	
## imprace:movies	-3.545976	1.689619	-2.099	0.03584	*
## imprace:shopping	-0.945740	0.988694	-0.957	0.33879	
## imprace:age_o	-0.085895	0.077685	-1.106	0.26886	
## tvsports:clubbing	1.325086	1.020763	1.298	0.19424	
## tvsports:movies	-1.858386	1.646502	-1.129	0.25903	

```
## tvsports:shopping      2.052988    1.097223    1.871    0.06133 .
## tvsports:age_o        -0.067017    0.081607   -0.821    0.41152
## clubbing:movies        0.036597    1.699280    0.022    0.98282
## clubbing:shopping     -0.499216    1.172573   -0.426    0.67029
## clubbing:age_o         0.104458    0.087209    1.198    0.23100
## movies:shopping       -0.113312    1.734800   -0.065    0.94792
## movies:age_o           0.024377    0.130500    0.187    0.85182
## shopping:age_o        -0.036149    0.086283   -0.419    0.67525
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 2495.3  on 2721  degrees of freedom
## Residual deviance: 1730.8  on 2616  degrees of freedom
##      (684 observations deleted due to missingness)
## AIC: 1942.8
##
## Number of Fisher Scoring iterations: 7
```

```
predict <- predict(m3, test_model_data, type = 'response')
# confusion matrix
table_mat <- table(test_model_data$match, predict > 0.5)
table_mat
```

```
##
##      FALSE TRUE
##  0    546   18
##  1     86   40
```

```
accuracy(table_mat)
```

```
## [1] 0.8492754
```

```
precision(table_mat)
```

```
## [1] 0.6896552
```

```
recall(table_mat)
```

```
## [1] 0.3174603
```

```
index <-
  predict[!is.na(predict)] %>%
  as.data.frame() %>%
  add_rownames() %>%
  pull(rowname)
```

```
t <-
```

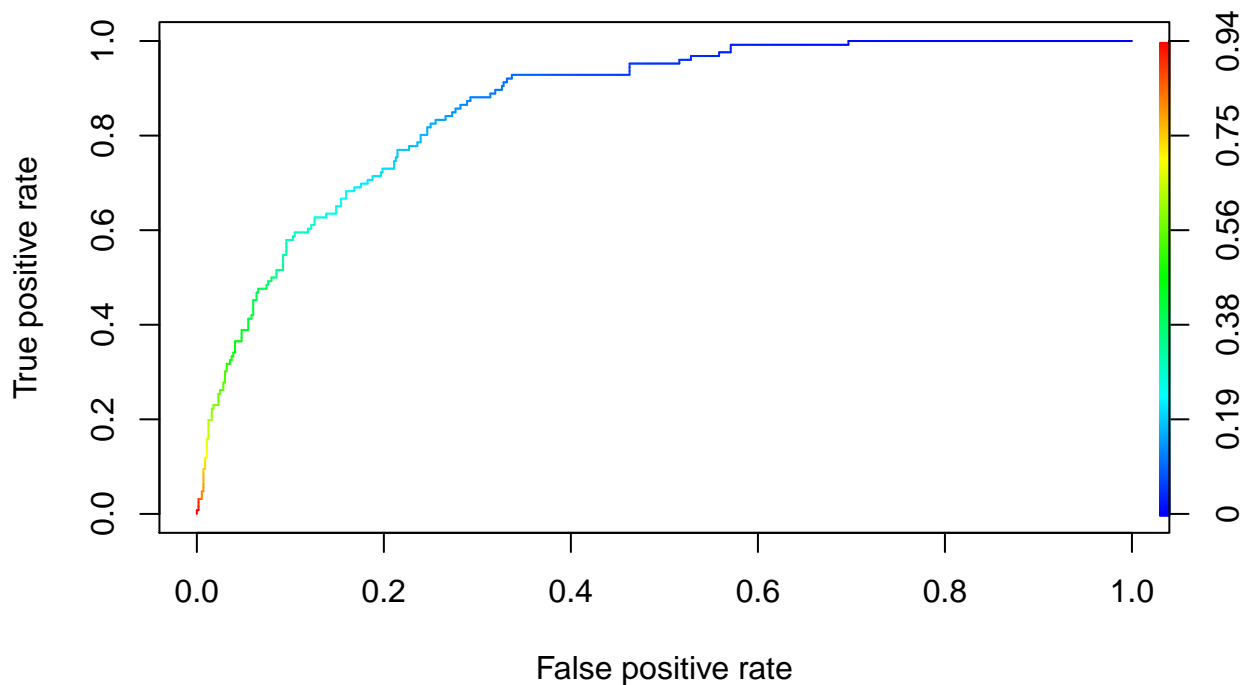


```

test_model_data %>%
  add_rownames() %>%
  filter(rowname %in% index)

ROCRpred <- prediction(predict[!is.na(predict)], t$match)
ROCRperf <- performance(ROCRpred, 'tpr', 'fpr')
plot(ROCRperf, colorize = TRUE, text.adj = c(-0.2, 1.7))

```



Stepwise AIC regression on model 2

```

train_model_data <-
  train_model_data %>%
  drop_na()

full_model = glm(match ~ . + .:., family = binomial(), data = train_model_data)

#summary(full_model)

null_model = glm(match ~ 1, family = binomial(), data = train_model_data)

```

```
#summary(null_model)
```

```
steps <-
```

```
  step(null_model, list(upper = full_model), direction = 'forward')
```

```
## Start: AIC=2497.31
```

```
## match ~ 1
```

```
##
```

	Df	Deviance	AIC
## + like	1	2183.0	2187.0
## + like_o	1	2208.4	2212.4
## + avg_partner_rating	1	2247.8	2251.8
## + prob	1	2310.4	2314.4
## + prob_o	1	2345.9	2349.9
## + fun	1	2458.0	2462.0
## + sinc	1	2467.8	2471.8
## + imprace	1	2477.5	2481.5
## + attr	1	2479.6	2483.6
## + clubbing	1	2480.2	2484.2
## <none>		2495.3	2497.3
## + age_o	1	2493.6	2497.6
## + shopping	1	2493.8	2497.8
## + movies	1	2494.0	2498.0
## + tvsports	1	2494.2	2498.2

```
##
```

```
## Step: AIC=2186.97
```

```
## match ~ like
```

```
##
```

	Df	Deviance	AIC
## + like_o	1	1949.5	1955.5
## + prob_o	1	2081.0	2087.0
## + prob	1	2145.8	2151.8
## + clubbing	1	2167.3	2173.3
## + imprace	1	2170.7	2176.7
## + avg_partner_rating	1	2171.3	2177.3
## + fun	1	2175.4	2181.4
## + sinc	1	2177.2	2183.2
## + tvsports	1	2180.1	2186.1
## + shopping	1	2180.3	2186.3
## + attr	1	2180.5	2186.5
## <none>		2183.0	2187.0
## + movies	1	2182.6	2188.6
## + age_o	1	2182.9	2188.9

```
##
```

```
## Step: AIC=1955.52
```

```
## match ~ like + like_o
```

```
##
```

	Df	Deviance	AIC
## + prob	1	1924.4	1932.4
## + prob_o	1	1934.0	1942.0
## + imprace	1	1939.9	1947.9
## + sinc	1	1940.1	1948.1
## + fun	1	1941.6	1949.6

```

## + clubbing          1  1943.0 1951.0
## + attr              1  1943.7 1951.7
## + avg_partner_rating 1  1944.2 1952.2
## + like:like_o       1  1945.5 1953.5
## + tvsports          1  1945.8 1953.8
## + shopping          1  1947.5 1955.5
## <none>              1949.5 1955.5
## + movies            1  1948.3 1956.3
## + age_o             1  1949.5 1957.5
##
## Step:  AIC=1932.4
## match ~ like + like_o + prob
##
##              Df Deviance   AIC
## + attr          1  1912.2 1922.2
## + prob_o        1  1912.3 1922.3
## + imprace       1  1914.8 1924.8
## + sinc          1  1915.2 1925.2
## + prob:like     1  1915.9 1925.9
## + fun           1  1916.4 1926.4
## + tvsports      1  1919.3 1929.3
## + shopping      1  1919.3 1929.3
## + like:like_o   1  1920.6 1930.6
## + clubbing      1  1921.1 1931.1
## <none>          1924.4 1932.4
## + avg_partner_rating 1  1923.0 1933.0
## + movies        1  1923.7 1933.7
## + age_o         1  1924.4 1934.4
## + prob:like_o   1  1924.4 1934.4
##
## Step:  AIC=1922.2
## match ~ like + like_o + prob + attr
##
##              Df Deviance   AIC
## + prob_o        1  1901.4 1913.4
## + prob:like     1  1902.8 1914.8
## + fun           1  1903.2 1915.2
## + imprace       1  1903.2 1915.2
## + sinc          1  1906.9 1918.9
## + tvsports      1  1907.5 1919.5
## + attr:like_o   1  1907.9 1919.9
## + shopping      1  1908.3 1920.3
## + like:like_o   1  1908.7 1920.7
## + clubbing      1  1908.8 1920.8
## + avg_partner_rating 1  1908.9 1920.9
## <none>          1912.2 1922.2
## + movies        1  1912.0 1924.0
## + attr:like     1  1912.0 1924.0
## + prob:like_o   1  1912.1 1924.1
## + age_o         1  1912.2 1924.2
## + attr:prob     1  1912.2 1924.2
##
## Step:  AIC=1913.44
## match ~ like + like_o + prob + attr + prob_o

```

```

##
##           Df Deviance    AIC
## + prob:like      1  1891.7 1905.7
## + imprace        1  1892.3 1906.3
## + fun            1  1893.7 1907.7
## + attr:like_o    1  1896.0 1910.0
## + tvsports       1  1896.1 1910.1
## + prob_o:like_o  1  1896.9 1910.9
## + sinc           1  1897.1 1911.1
## + like:like_o    1  1897.6 1911.6
## + shopping       1  1898.1 1912.1
## + clubbing       1  1898.2 1912.2
## + avg_partner_rating 1  1898.3 1912.3
## <none>          1901.4 1913.4
## + prob_o:like    1  1901.1 1915.1
## + prob:like_o    1  1901.2 1915.2
## + movies         1  1901.3 1915.3
## + age_o          1  1901.3 1915.3
## + attr:like      1  1901.3 1915.3
## + attr:prob_o    1  1901.4 1915.4
## + prob:prob_o    1  1901.4 1915.4
## + attr:prob      1  1901.4 1915.4
##
## Step:  AIC=1905.69
## match ~ like + like_o + prob + attr + prob_o + like:prob
##
##           Df Deviance    AIC
## + imprace        1  1881.6 1897.6
## + fun            1  1884.2 1900.2
## + attr:like_o    1  1885.5 1901.5
## + tvsports       1  1886.5 1902.5
## + sinc           1  1887.1 1903.1
## + like:like_o    1  1887.6 1903.6
## + prob_o:like_o  1  1887.8 1903.8
## + avg_partner_rating 1  1888.2 1904.2
## + clubbing       1  1888.3 1904.3
## + shopping       1  1888.6 1904.6
## <none>          1891.7 1905.7
## + prob:like_o    1  1891.3 1907.3
## + movies         1  1891.5 1907.5
## + prob_o:like    1  1891.5 1907.5
## + age_o          1  1891.6 1907.6
## + attr:like      1  1891.6 1907.6
## + attr:prob_o    1  1891.7 1907.7
## + attr:prob      1  1891.7 1907.7
## + prob:prob_o    1  1891.7 1907.7
##
## Step:  AIC=1897.63
## match ~ like + like_o + prob + attr + prob_o + imprace + like:prob
##
##           Df Deviance    AIC
## + fun            1  1873.9 1891.9
## + attr:like_o    1  1875.5 1893.5
## + avg_partner_rating 1  1876.6 1894.6

```

```

## + tvsports          1  1877.2 1895.2
## + sinc              1  1877.3 1895.3
## + like:like_o       1  1877.4 1895.4
## + prob_o:like_o     1  1877.6 1895.6
## + clubbing          1  1879.6 1897.6
## <none>              1881.6 1897.6
## + shopping          1  1880.4 1898.4
## + like:imprace      1  1880.8 1898.8
## + like_o:imprace    1  1881.2 1899.2
## + prob:like_o       1  1881.4 1899.4
## + prob:imprace      1  1881.5 1899.5
## + age_o             1  1881.5 1899.5
## + prob_o:like       1  1881.5 1899.5
## + prob_o:imprace    1  1881.6 1899.6
## + attr:like         1  1881.6 1899.6
## + attr:prob_o       1  1881.6 1899.6
## + movies            1  1881.6 1899.6
## + attr:prob         1  1881.6 1899.6
## + prob:prob_o       1  1881.6 1899.6
## + attr:imprace      1  1881.6 1899.6
##
## Step:  AIC=1891.86
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##      like:prob
##
##              Df Deviance    AIC
## + attr:like_o      1  1867.8 1887.8
## + avg_partner_rating 1  1869.1 1889.1
## + prob_o:like_o    1  1869.3 1889.3
## + like:like_o      1  1869.7 1889.7
## + tvsports         1  1869.7 1889.7
## + attr:fun         1  1871.8 1891.8
## <none>             1873.9 1891.9
## + sinc            1  1872.1 1892.1
## + clubbing        1  1872.1 1892.1
## + shopping        1  1872.7 1892.7
## + like:imprace    1  1873.0 1893.0
## + fun:like        1  1873.3 1893.3
## + like_o:imprace  1  1873.4 1893.4
## + age_o           1  1873.5 1893.5
## + fun:imprace     1  1873.6 1893.6
## + prob:like_o     1  1873.7 1893.7
## + prob:imprace    1  1873.7 1893.7
## + attr:prob_o     1  1873.8 1893.8
## + attr:like       1  1873.8 1893.8
## + prob_o:like     1  1873.8 1893.8
## + prob_o:imprace  1  1873.8 1893.8
## + fun:prob        1  1873.8 1893.8
## + fun:like_o      1  1873.8 1893.8
## + prob:prob_o     1  1873.8 1893.8
## + movies          1  1873.8 1893.8
## + attr:prob       1  1873.8 1893.8
## + attr:imprace    1  1873.8 1893.8
## + fun:prob_o      1  1873.8 1893.8

```

```

##
## Step: AIC=1887.78
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##      like:prob + like_o:attr
##
##           Df Deviance    AIC
## + avg_partner_rating 1  1862.6 1884.6
## + prob_o:like_o      1  1862.8 1884.8
## + tvsports           1  1863.6 1885.6
## + like:like_o        1  1864.5 1886.5
## <none>               1  1867.8 1887.8
## + sinc               1  1865.9 1887.9
## + clubbing           1  1866.1 1888.1
## + attr:fun           1  1866.4 1888.4
## + attr:prob_o        1  1866.5 1888.5
## + shopping           1  1866.6 1888.6
## + like:imprace       1  1867.1 1889.1
## + fun:like           1  1867.2 1889.2
## + fun:imprace        1  1867.4 1889.4
## + age_o              1  1867.5 1889.5
## + like_o:imprace     1  1867.5 1889.5
## + prob:imprace       1  1867.6 1889.6
## + prob_o:like        1  1867.7 1889.7
## + attr:prob          1  1867.7 1889.7
## + fun:like_o         1  1867.7 1889.7
## + fun:prob           1  1867.8 1889.8
## + attr:like          1  1867.8 1889.8
## + prob:prob_o        1  1867.8 1889.8
## + prob_o:imprace     1  1867.8 1889.8
## + movies             1  1867.8 1889.8
## + fun:prob_o         1  1867.8 1889.8
## + attr:imprace       1  1867.8 1889.8
## + prob:like_o        1  1867.8 1889.8
##
## Step: AIC=1884.56
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##      avg_partner_rating + like:prob + like_o:attr
##
##           Df Deviance    AIC
## + like:avg_partner_rating 1  1848.6 1872.6
## + tvsports                1  1857.7 1881.7
## + prob_o:like_o           1  1857.8 1881.8
## + like:like_o             1  1859.3 1883.3
## + like_o:avg_partner_rating 1  1859.6 1883.6
## <none>                    1  1862.6 1884.6
## + clubbing                1  1860.7 1884.7
## + sinc                    1  1861.1 1885.1
## + attr:fun                1  1861.1 1885.1
## + shopping                1  1861.2 1885.2
## + attr:prob_o             1  1861.3 1885.3
## + avg_partner_rating:imprace 1  1861.5 1885.5
## + like:imprace            1  1861.6 1885.6
## + prob:avg_partner_rating  1  1861.9 1885.9
## + fun:like                1  1862.1 1886.1

```

```

## + fun:imprace      1  1862.2 1886.2
## + age_o           1  1862.2 1886.2
## + like_o:imprace   1  1862.2 1886.2
## + fun:avg_partner_rating 1  1862.4 1886.4
## + prob:imprace     1  1862.4 1886.4
## + prob_o:avg_partner_rating 1  1862.4 1886.4
## + prob_o:like      1  1862.5 1886.5
## + fun:like_o       1  1862.5 1886.5
## + movies          1  1862.5 1886.5
## + fun:prob        1  1862.5 1886.5
## + attr:prob       1  1862.5 1886.5
## + attr:avg_partner_rating 1  1862.5 1886.5
## + prob:like_o      1  1862.6 1886.6
## + prob_o:imprace   1  1862.6 1886.6
## + fun:prob_o       1  1862.6 1886.6
## + attr:like        1  1862.6 1886.6
## + attr:imprace     1  1862.6 1886.6
## + prob:prob_o      1  1862.6 1886.6
##
## Step:  AIC=1872.57
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##         avg_partner_rating + like:prob + like_o:attr + like:avg_partner_rating
##
##               Df Deviance    AIC
## + prob_o:like_o      1  1843.5 1869.5
## + tvsports          1  1844.0 1870.0
## + like:like_o        1  1844.6 1870.6
## + like_o:avg_partner_rating 1  1845.4 1871.4
## + clubbing          1  1846.3 1872.3
## + attr:fun          1  1846.5 1872.5
## <none>              1848.6 1872.6
## + sinc              1  1846.8 1872.8
## + shopping          1  1847.1 1873.1
## + attr:prob_o        1  1847.3 1873.3
## + fun:avg_partner_rating 1  1847.5 1873.5
## + like:imprace       1  1847.8 1873.8
## + avg_partner_rating:imprace 1  1848.0 1874.0
## + age_o             1  1848.2 1874.2
## + like_o:imprace     1  1848.2 1874.2
## + fun:imprace        1  1848.3 1874.3
## + fun:like           1  1848.3 1874.3
## + attr:avg_partner_rating 1  1848.4 1874.4
## + movies            1  1848.4 1874.4
## + attr:like          1  1848.4 1874.4
## + prob_o:avg_partner_rating 1  1848.5 1874.5
## + prob:imprace       1  1848.5 1874.5
## + prob:avg_partner_rating 1  1848.5 1874.5
## + prob_o:like        1  1848.5 1874.5
## + fun:like_o         1  1848.5 1874.5
## + attr:imprace       1  1848.5 1874.5
## + fun:prob           1  1848.5 1874.5
## + prob:like_o        1  1848.5 1874.5
## + attr:prob          1  1848.6 1874.6
## + prob:prob_o        1  1848.6 1874.6

```

```

## + fun:prob_o          1  1848.6 1874.6
## + prob_o:imprace      1  1848.6 1874.6
##
## Step:  AIC=1869.48
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##      avg_partner_rating + like:prob + like_o:attr + like:avg_partner_rating +
##      like_o:prob_o
##
##              Df Deviance    AIC
## + tvsports      1  1839.0 1867.0
## + like:like_o    1  1839.1 1867.1
## + like_o:avg_partner_rating 1  1839.9 1867.9
## + clubbing      1  1841.4 1869.4
## <none>          1  1843.5 1869.5
## + sinc          1  1841.6 1869.6
## + attr:fun       1  1841.7 1869.7
## + shopping      1  1842.0 1870.0
## + attr:prob_o    1  1842.3 1870.3
## + fun:avg_partner_rating 1  1842.6 1870.6
## + like:imprace   1  1842.7 1870.7
## + avg_partner_rating:imprace 1  1842.8 1870.8
## + fun:imprace    1  1843.0 1871.0
## + like_o:imprace 1  1843.1 1871.1
## + fun:like       1  1843.2 1871.2
## + attr:avg_partner_rating 1  1843.2 1871.2
## + prob:imprace   1  1843.3 1871.3
## + movies        1  1843.3 1871.3
## + attr:like      1  1843.3 1871.3
## + age_o         1  1843.3 1871.3
## + prob_o:like    1  1843.3 1871.3
## + prob_o:avg_partner_rating 1  1843.3 1871.3
## + fun:like_o     1  1843.4 1871.4
## + prob:avg_partner_rating 1  1843.4 1871.4
## + attr:imprace   1  1843.4 1871.4
## + fun:prob_o     1  1843.5 1871.5
## + fun:prob       1  1843.5 1871.5
## + prob_o:imprace 1  1843.5 1871.5
## + prob:like_o    1  1843.5 1871.5
## + prob:prob_o    1  1843.5 1871.5
## + attr:prob      1  1843.5 1871.5
##
## Step:  AIC=1867.01
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##      avg_partner_rating + tvsports + like:prob + like_o:attr +
##      like:avg_partner_rating + like_o:prob_o
##
##              Df Deviance    AIC
## + like:like_o    1  1834.5 1864.5
## + clubbing      1  1835.1 1865.1
## + like_o:avg_partner_rating 1  1835.3 1865.3
## + prob:tvsports  1  1835.5 1865.5
## + attr:tvsports  1  1836.7 1866.7
## + sinc          1  1836.9 1866.9
## <none>          1  1839.0 1867.0

```



```

## + attr:fun          1  1837.0 1867.0
## + shopping          1  1837.8 1867.8
## + attr:prob_o       1  1837.9 1867.9
## + fun:avg_partner_rating 1  1838.0 1868.0
## + fun:tvsports      1  1838.2 1868.2
## + like:imprace      1  1838.5 1868.5
## + like:tvsports     1  1838.5 1868.5
## + avg_partner_rating:imprace 1  1838.5 1868.5
## + imprace:tvsports  1  1838.5 1868.5
## + fun:imprace       1  1838.6 1868.6
## + like_o:imprace    1  1838.6 1868.6
## + fun:like          1  1838.7 1868.7
## + attr:avg_partner_rating 1  1838.7 1868.7
## + movies            1  1838.7 1868.7
## + attr:like         1  1838.8 1868.8
## + prob:imprace      1  1838.8 1868.8
## + age_o             1  1838.8 1868.8
## + avg_partner_rating:tvsports 1  1838.9 1868.9
## + prob_o:avg_partner_rating 1  1838.9 1868.9
## + prob_o:like       1  1838.9 1868.9
## + attr:imprace      1  1839.0 1869.0
## + fun:like_o        1  1839.0 1869.0
## + prob:like_o       1  1839.0 1869.0
## + fun:prob_o        1  1839.0 1869.0
## + like_o:tvsports   1  1839.0 1869.0
## + prob_o:imprace    1  1839.0 1869.0
## + prob:avg_partner_rating 1  1839.0 1869.0
## + prob_o:tvsports   1  1839.0 1869.0
## + attr:prob         1  1839.0 1869.0
## + prob:prob_o       1  1839.0 1869.0
## + fun:prob          1  1839.0 1869.0
##
## Step:  AIC=1864.51
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##         avg_partner_rating + tvsports + like:prob + like_o:attr +
##         like:avg_partner_rating + like_o:prob_o + like:like_o
##
##
##           Df Deviance    AIC
## + clubbing          1  1830.4 1862.4
## + prob:tvsports     1  1831.1 1863.1
## + attr:fun          1  1832.2 1864.2
## + attr:tvsports     1  1832.2 1864.2
## + sinc              1  1832.4 1864.4
## + <none>             1834.5 1864.5
## + attr:prob_o       1  1833.2 1865.2
## + shopping          1  1833.3 1865.3
## + prob_o:like       1  1833.3 1865.3
## + fun:avg_partner_rating 1  1833.5 1865.5
## + fun:tvsports      1  1833.7 1865.7
## + like:tvsports     1  1833.9 1865.9
## + prob:like_o       1  1834.0 1866.0
## + imprace:tvsports  1  1834.0 1866.0
## + like_o:imprace    1  1834.0 1866.0
## + like_o:avg_partner_rating 1  1834.1 1866.1

```

```

## + fun:imprace      1  1834.1 1866.1
## + avg_partner_rating:imprace 1  1834.1 1866.1
## + like:imprace     1  1834.2 1866.2
## + movies           1  1834.2 1866.2
## + age_o            1  1834.3 1866.3
## + fun:like         1  1834.3 1866.3
## + avg_partner_rating:tvsports 1  1834.4 1866.4
## + attr:avg_partner_rating 1  1834.4 1866.4
## + prob:prob_o      1  1834.4 1866.4
## + prob:imprace     1  1834.4 1866.4
## + attr:imprace     1  1834.4 1866.4
## + attr:like        1  1834.4 1866.4
## + prob_o:imprace   1  1834.5 1866.5
## + prob_o:tvsports  1  1834.5 1866.5
## + prob_o:avg_partner_rating 1  1834.5 1866.5
## + fun:prob_o       1  1834.5 1866.5
## + attr:prob        1  1834.5 1866.5
## + fun:like_o       1  1834.5 1866.5
## + like_o:tvsports  1  1834.5 1866.5
## + fun:prob         1  1834.5 1866.5
## + prob:avg_partner_rating 1  1834.5 1866.5
##
## Step:  AIC=1862.43
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##         avg_partner_rating + tvsports + clubbing + like:prob + like_o:attr +
##         like:avg_partner_rating + like_o:prob_o + like:like_o
##
##               Df Deviance    AIC
## + fun:clubbing      1  1825.1 1859.1
## + attr:clubbing     1  1825.3 1859.3
## + prob:tvsports     1  1826.8 1860.8
## + attr:tvsports     1  1828.0 1862.0
## + attr:fun          1  1828.1 1862.1
## <none>              1830.4 1862.4
## + tvsports:clubbing 1  1828.6 1862.6
## + shopping          1  1828.7 1862.7
## + sinc              1  1828.9 1862.9
## + attr:prob_o       1  1829.1 1863.1
## + like:clubbing     1  1829.3 1863.3
## + prob_o:like       1  1829.4 1863.4
## + fun:avg_partner_rating 1  1829.5 1863.5
## + prob:clubbing     1  1829.5 1863.5
## + fun:tvsports      1  1829.6 1863.6
## + movies            1  1829.8 1863.8
## + like_o:avg_partner_rating 1  1829.8 1863.8
## + avg_partner_rating:imprace 1  1829.8 1863.8
## + imprace:tvsports  1  1829.8 1863.8
## + prob:like_o       1  1829.9 1863.9
## + avg_partner_rating:clubbing 1  1830.0 1864.0
## + like:tvsports     1  1830.0 1864.0
## + like:imprace      1  1830.0 1864.0
## + like_o:imprace    1  1830.0 1864.0
## + fun:imprace       1  1830.1 1864.1
## + age_o             1  1830.1 1864.1

```

```

## + prob:imprace          1  1830.2 1864.2
## + attr:avg_partner_rating 1  1830.2 1864.2
## + fun:like              1  1830.2 1864.2
## + imprace:clubbing      1  1830.3 1864.3
## + attr:like             1  1830.3 1864.3
## + attr:imprace         1  1830.3 1864.3
## + prob:prob_o          1  1830.4 1864.4
## + like_o:clubbing      1  1830.4 1864.4
## + avg_partner_rating:tvsports 1  1830.4 1864.4
## + prob_o:tvsports      1  1830.4 1864.4
## + fun:prob_o           1  1830.4 1864.4
## + fun:like_o           1  1830.4 1864.4
## + prob_o:imprace       1  1830.4 1864.4
## + prob_o:avg_partner_rating 1  1830.4 1864.4
## + attr:prob            1  1830.4 1864.4
## + prob:avg_partner_rating 1  1830.4 1864.4
## + prob_o:clubbing      1  1830.4 1864.4
## + like_o:tvsports      1  1830.4 1864.4
## + fun:prob             1  1830.4 1864.4
##
## Step: AIC=1859.06
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##         avg_partner_rating + tvsports + clubbing + like:prob + like_o:attr +
##         like:avg_partner_rating + like_o:prob_o + like:like_o + fun:clubbing
##
##               Df Deviance    AIC
## + attr:clubbing      1  1820.4 1856.4
## + prob:tvsports      1  1821.4 1857.4
## + attr:tvsports      1  1822.4 1858.4
## + attr:fun           1  1822.8 1858.8
## <none>                1825.1 1859.1
## + like:clubbing      1  1823.1 1859.1
## + shopping           1  1823.3 1859.3
## + tvsports:clubbing  1  1823.4 1859.4
## + fun:avg_partner_rating 1  1823.5 1859.5
## + attr:prob_o        1  1823.5 1859.5
## + sinc               1  1823.9 1859.9
## + prob_o:like        1  1823.9 1859.9
## + fun:tvsports       1  1824.2 1860.2
## + avg_partner_rating:imprace 1  1824.3 1860.3
## + prob:clubbing      1  1824.3 1860.3
## + movies             1  1824.5 1860.5
## + prob:like_o        1  1824.5 1860.5
## + imprace:tvsports   1  1824.5 1860.5
## + like_o:imprace     1  1824.5 1860.5
## + like_o:avg_partner_rating 1  1824.5 1860.5
## + like:imprace       1  1824.6 1860.6
## + like:tvsports      1  1824.6 1860.6
## + age_o              1  1824.7 1860.7
## + prob:imprace       1  1824.8 1860.8
## + imprace:clubbing   1  1824.8 1860.8
## + avg_partner_rating:clubbing 1  1824.9 1860.9
## + attr:imprace       1  1824.9 1860.9
## + attr:avg_partner_rating 1  1824.9 1860.9

```

```

## + attr:like          1  1824.9 1860.9
## + like_o:clubbing    1  1824.9 1860.9
## + fun:imprace        1  1825.0 1861.0
## + fun:like_o         1  1825.0 1861.0
## + prob_o:tvsports    1  1825.0 1861.0
## + fun:like           1  1825.0 1861.0
## + avg_partner_rating:tvsports 1  1825.0 1861.0
## + prob:prob_o        1  1825.0 1861.0
## + prob_o:imprace     1  1825.0 1861.0
## + prob:avg_partner_rating 1  1825.0 1861.0
## + attr:prob          1  1825.0 1861.0
## + fun:prob           1  1825.0 1861.0
## + prob_o:avg_partner_rating 1  1825.0 1861.0
## + fun:prob_o         1  1825.1 1861.1
## + like_o:tvsports    1  1825.1 1861.1
## + prob_o:clubbing    1  1825.1 1861.1
##
## Step:  AIC=1856.39
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##         avg_partner_rating + tvsports + clubbing + like:prob + like_o:attr +
##         like:avg_partner_rating + like_o:prob_o + like:like_o + fun:clubbing +
##         attr:clubbing
##
##
##           Df Deviance    AIC
## + prob:tvsports      1  1816.4 1854.4
## + attr:tvsports      1  1816.9 1854.9
## <none>                1820.4 1856.4
## + attr:fun           1  1818.5 1856.5
## + shopping           1  1818.6 1856.6
## + attr:prob_o        1  1818.7 1856.7
## + fun:avg_partner_rating 1  1818.8 1856.8
## + tvsports:clubbing  1  1819.0 1857.0
## + like:clubbing      1  1819.0 1857.0
## + sinc               1  1819.1 1857.1
## + prob_o:like        1  1819.2 1857.2
## + avg_partner_rating:imprace 1  1819.4 1857.4
## + fun:tvsports       1  1819.4 1857.4
## + movies             1  1819.7 1857.7
## + prob:like_o        1  1819.8 1857.8
## + like_o:imprace     1  1819.8 1857.8
## + attr:imprace       1  1819.8 1857.8
## + like_o:avg_partner_rating 1  1819.9 1857.9
## + imprace:tvsports   1  1819.9 1857.9
## + like:imprace       1  1819.9 1857.9
## + like:tvsports      1  1819.9 1857.9
## + age_o              1  1820.0 1858.0
## + imprace:clubbing   1  1820.0 1858.0
## + prob:imprace       1  1820.1 1858.1
## + prob:clubbing      1  1820.2 1858.2
## + attr:avg_partner_rating 1  1820.3 1858.3
## + avg_partner_rating:tvsports 1  1820.3 1858.3
## + like_o:clubbing    1  1820.3 1858.3
## + fun:imprace        1  1820.3 1858.3
## + fun:like_o         1  1820.3 1858.3

```

```

## + attr:like 1 1820.3 1858.3
## + avg_partner_rating:clubbing 1 1820.3 1858.3
## + prob_o:imprace 1 1820.3 1858.3
## + prob_o:clubbing 1 1820.4 1858.4
## + fun:like 1 1820.4 1858.4
## + attr:prob 1 1820.4 1858.4
## + prob:prob_o 1 1820.4 1858.4
## + prob_o:tvsports 1 1820.4 1858.4
## + fun:prob 1 1820.4 1858.4
## + prob:avg_partner_rating 1 1820.4 1858.4
## + like_o:tvsports 1 1820.4 1858.4
## + fun:prob_o 1 1820.4 1858.4
## + prob_o:avg_partner_rating 1 1820.4 1858.4
##
## Step: AIC=1854.44
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
## avg_partner_rating + tvsports + clubbing + like:prob + like_o:attr +
## like:avg_partner_rating + like_o:prob_o + like:like_o + fun:clubbing +
## attr:clubbing + prob:tvsports
##
## Df Deviance AIC
## + attr:tvsports 1 1813.8 1853.8
## + like:tvsports 1 1814.2 1854.2
## <none> 1816.4 1854.4
## + tvsports:clubbing 1 1814.4 1854.4
## + attr:fun 1 1814.7 1854.7
## + attr:prob_o 1 1814.8 1854.8
## + sinc 1 1814.9 1854.9
## + fun:avg_partner_rating 1 1815.0 1855.0
## + shopping 1 1815.0 1855.0
## + avg_partner_rating:tvsports 1 1815.1 1855.1
## + prob_o:like 1 1815.3 1855.3
## + avg_partner_rating:imprace 1 1815.3 1855.3
## + like:clubbing 1 1815.5 1855.5
## + fun:tvsports 1 1815.5 1855.5
## + movies 1 1815.8 1855.8
## + prob:like_o 1 1815.8 1855.8
## + like_o:avg_partner_rating 1 1815.9 1855.9
## + like:imprace 1 1815.9 1855.9
## + attr:imprace 1 1815.9 1855.9
## + imprace:tvsports 1 1815.9 1855.9
## + like_o:imprace 1 1815.9 1855.9
## + prob:clubbing 1 1815.9 1855.9
## + age_o 1 1816.0 1856.0
## + imprace:clubbing 1 1816.0 1856.0
## + prob:imprace 1 1816.1 1856.1
## + avg_partner_rating:clubbing 1 1816.3 1856.3
## + attr:avg_partner_rating 1 1816.3 1856.3
## + fun:imprace 1 1816.3 1856.3
## + like_o:clubbing 1 1816.4 1856.4
## + fun:like_o 1 1816.4 1856.4
## + prob:avg_partner_rating 1 1816.4 1856.4
## + attr:like 1 1816.4 1856.4
## + prob_o:imprace 1 1816.4 1856.4

```

```

## + prob_o:tvsports          1  1816.4 1856.4
## + attr:prob                1  1816.4 1856.4
## + prob:prob_o              1  1816.4 1856.4
## + prob_o:clubbing          1  1816.4 1856.4
## + fun:like                  1  1816.4 1856.4
## + fun:prob                 1  1816.4 1856.4
## + like_o:tvsports          1  1816.4 1856.4
## + fun:prob_o               1  1816.4 1856.4
## + prob_o:avg_partner_rating 1  1816.4 1856.4
##
## Step: AIC=1853.8
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##      avg_partner_rating + tvsports + clubbing + like:prob + like_o:attr +
##      like:avg_partner_rating + like_o:prob_o + like:like_o + fun:clubbing +
##      attr:clubbing + prob:tvsports + attr:tvsports
##
##              Df Deviance   AIC
## + tvsports:clubbing      1  1811.4 1853.4
## <none>                    1813.8 1853.8
## + attr:prob_o            1  1812.2 1854.2
## + sinc                    1  1812.2 1854.2
## + avg_partner_rating:tvsports 1  1812.4 1854.4
## + like:tvsports          1  1812.4 1854.4
## + attr:fun                1  1812.4 1854.4
## + fun:avg_partner_rating  1  1812.5 1854.5
## + shopping                1  1812.5 1854.5
## + prob_o:like             1  1812.6 1854.6
## + avg_partner_rating:imprace 1  1812.8 1854.8
## + like:clubbing          1  1812.9 1854.9
## + prob:like_o            1  1813.1 1855.1
## + attr:imprace           1  1813.1 1855.1
## + fun:tvsports           1  1813.1 1855.1
## + like_o:imprace         1  1813.2 1855.2
## + movies                  1  1813.2 1855.2
## + imprace:tvsports       1  1813.2 1855.2
## + like:imprace           1  1813.3 1855.3
## + prob:clubbing          1  1813.3 1855.3
## + like_o:avg_partner_rating 1  1813.4 1855.4
## + imprace:clubbing       1  1813.4 1855.4
## + age_o                   1  1813.4 1855.4
## + prob:imprace           1  1813.5 1855.5
## + like_o:clubbing        1  1813.7 1855.7
## + fun:imprace            1  1813.7 1855.7
## + fun:like_o             1  1813.7 1855.7
## + prob_o:imprace         1  1813.7 1855.7
## + avg_partner_rating:clubbing 1  1813.8 1855.8
## + prob:prob_o            1  1813.8 1855.8
## + prob:avg_partner_rating 1  1813.8 1855.8
## + prob_o:tvsports        1  1813.8 1855.8
## + fun:like                1  1813.8 1855.8
## + attr:prob              1  1813.8 1855.8
## + attr:like              1  1813.8 1855.8
## + attr:avg_partner_rating 1  1813.8 1855.8
## + fun:prob               1  1813.8 1855.8

```

```

## + fun:prob_o          1  1813.8 1855.8
## + prob_o:avg_partner_rating 1  1813.8 1855.8
## + prob_o:clubbing     1  1813.8 1855.8
## + like_o:tvsports     1  1813.8 1855.8
##
## Step: AIC=1853.41
## match ~ like + like_o + prob + attr + prob_o + imprace + fun +
##      avg_partner_rating + tvsports + clubbing + like:prob + like_o:attr +
##      like:avg_partner_rating + like_o:prob_o + like:like_o + fun:clubbing +
##      attr:clubbing + prob:tvsports + attr:tvsports + tvsports:clubbing
##
##              Df Deviance    AIC
## <none>              1811.4 1853.4
## + attr:prob_o          1  1809.5 1853.5
## + avg_partner_rating:tvsports 1  1809.7 1853.7
## + like:tvsports        1  1809.8 1853.8
## + sinc                 1  1809.8 1853.8
## + attr:fun             1  1810.1 1854.1
## + prob_o:like          1  1810.1 1854.1
## + fun:avg_partner_rating 1  1810.1 1854.1
## + shopping            1  1810.2 1854.2
## + avg_partner_rating:imprace 1  1810.3 1854.3
## + fun:tvsports         1  1810.4 1854.4
## + like:clubbing        1  1810.5 1854.5
## + attr:imprace         1  1810.7 1854.7
## + prob:like_o          1  1810.7 1854.7
## + movies               1  1810.7 1854.7
## + like_o:avg_partner_rating 1  1810.8 1854.8
## + like:imprace         1  1810.9 1854.9
## + imprace:clubbing     1  1810.9 1854.9
## + like_o:imprace       1  1810.9 1854.9
## + age_o                1  1811.0 1855.0
## + prob:imprace         1  1811.0 1855.0
## + prob:clubbing        1  1811.0 1855.0
## + imprace:tvsports     1  1811.3 1855.3
## + fun:imprace          1  1811.3 1855.3
## + prob:avg_partner_rating 1  1811.3 1855.3
## + fun:like_o           1  1811.4 1855.4
## + like_o:clubbing      1  1811.4 1855.4
## + prob_o:imprace       1  1811.4 1855.4
## + avg_partner_rating:clubbing 1  1811.4 1855.4
## + attr:like            1  1811.4 1855.4
## + attr:prob            1  1811.4 1855.4
## + prob:prob_o          1  1811.4 1855.4
## + fun:prob_o           1  1811.4 1855.4
## + prob_o:tvsports      1  1811.4 1855.4
## + fun:like             1  1811.4 1855.4
## + fun:prob             1  1811.4 1855.4
## + like_o:tvsports      1  1811.4 1855.4
## + prob_o:clubbing      1  1811.4 1855.4
## + prob_o:avg_partner_rating 1  1811.4 1855.4
## + attr:avg_partner_rating 1  1811.4 1855.4

```

```

best_model <-
  glm(formula = match ~ like + like_o + prob + attr + imprace +
      prob_o + fun + tvsports + avg_partner_rating + clubbing +
      movies + like:prob + like_o:attr + like:avg_partner_rating +
      like:like_o + like_o:prob_o + attr:tvsports + fun:clubbing +
      attr:clubbing + imprace:movies + like:movies + clubbing:movies +
      attr:prob_o + tvsports:clubbing + prob:tvsports + tvsports:avg_partner_rating +
      like:clubbing, family = binomial(), data = train_model_data)

predict <- predict(m2, test_model_data, type = 'response')
# confusion matrix
table_mat <- table(test_model_data$match, predict > 0.5)
table_mat

##
##      FALSE TRUE
##    0    549   15
##    1     95   31

accuracy(table_mat)

## [1] 0.8405797

precision(table_mat)

## [1] 0.673913

recall(table_mat)

## [1] 0.2460317

index <-
  predict[!is.na(predict)] %>%
  as.data.frame() %>%
  add_rownames() %>%
  pull(rowname)

summary(best_model)

##
## Call:
## glm(formula = match ~ like + like_o + prob + attr + imprace +
##      prob_o + fun + tvsports + avg_partner_rating + clubbing +
##      movies + like:prob + like_o:attr + like:avg_partner_rating +
##      like:like_o + like_o:prob_o + attr:tvsports + fun:clubbing +
##      attr:clubbing + imprace:movies + like:movies + clubbing:movies +
##      attr:prob_o + tvsports:clubbing + prob:tvsports + tvsports:avg_partner_rating +

```



```
##      like:clubbing, family = binomial(), data = train_model_data)
##
## Deviance Residuals:
##      Min        1Q      Median        3Q        Max
## -2.32202  -0.56597  -0.27440  -0.07125   3.03377
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -11.9769     3.3388  -3.587 0.000334 ***
## like           7.8042     3.7413   2.086 0.036984 *
## like_o        -4.1701     3.0942  -1.348 0.177754
## prob          4.5629     1.8108   2.520 0.011743 *
## attr         -7.6546    11.1708  -0.685 0.493196
## imprace       2.2230     1.1231   1.979 0.047788 *
## prob_o        6.2972     2.1720   2.899 0.003741 **
## fun          -6.4612     5.9831  -1.080 0.280183
## tvsports     -5.0877     2.2382  -2.273 0.023020 *
## avg_partner_rating 10.8923     2.9731   3.664 0.000249 ***
## clubbing       1.3230     2.5739   0.514 0.607249
## movies       -3.5385     2.1326  -1.659 0.097069 .
## like:prob     -2.3876     2.2707  -1.051 0.293031
## like_o:attr   42.3154    13.8853   3.047 0.002308 **
## like:avg_partner_rating -14.3853     3.7821  -3.804 0.000143 ***
## like:like_o    6.3184     3.1729   1.991 0.046440 *
## like_o:prob_o  -4.0751     1.8991  -2.146 0.031890 *
## attr:tvsports 13.5330     7.7181   1.753 0.079530 .
## fun:clubbing  21.4531     9.1531   2.344 0.019088 *
## attr:clubbing -18.5520     8.3521  -2.221 0.026335 *
## imprace:movies -3.6129     1.3858  -2.607 0.009130 **
## like:movies    5.6808     2.6073   2.179 0.029348 *
## clubbing:movies 0.5148     1.3284   0.388 0.698338
## attr:prob_o   -13.9883    10.0082  -1.398 0.162208
## tvsports:clubbing 1.5868     0.9038   1.756 0.079121 .
## prob:tvsports -2.6118     1.2600  -2.073 0.038184 *
## tvsports:avg_partner_rating 4.0675     2.3560   1.726 0.084275 .
## like:clubbing -3.3326     2.0098  -1.658 0.097277 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 2495.3  on 2721  degrees of freedom
## Residual deviance: 1793.9  on 2694  degrees of freedom
## AIC: 1849.9
##
## Number of Fisher Scoring iterations: 7
```

```
t <-
  test_model_data %>%
  add_rownames() %>%
  filter(rowname %in% index)
```

```
ROCpred <- prediction(predict[!is.na(predict)], t$match)
```

```
ROCperf <- performance(ROCpred, 'tpr', 'fpr')  
plot(ROCperf, colorize = TRUE, text.adj = c(-0.2, 1.7))
```

