

Customer Purchase Behavior

2024-07-22

```
library(caret)
```

```
## Warning: package 'caret' was built under R version 4.4.1
## Loading required package: ggplot2
## Warning: package 'ggplot2' was built under R version 4.4.1
## Loading required package: lattice
```

```
library(randomForest)
```

```
## Warning: package 'randomForest' was built under R version 4.4.1
## randomForest 4.7-1.1
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:ggplot2':
##
##     margin
```

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 4.4.1
## Warning: package 'tibble' was built under R version 4.4.1
## Warning: package 'tidyr' was built under R version 4.4.1
## Warning: package 'readr' was built under R version 4.4.1
## Warning: package 'purrr' was built under R version 4.4.1
## Warning: package 'dplyr' was built under R version 4.4.1
## Warning: package 'stringr' was built under R version 4.4.1
## Warning: package 'forcats' was built under R version 4.4.1
## Warning: package 'lubridate' was built under R version 4.4.1

## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr    1.5.1
## v lubridate  1.9.3      v tibble     3.2.1
## v purrr      1.0.2      v tidyr      1.3.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::combine()      masks randomForest::combine()
## x dplyr::filter()       masks stats::filter()
## x dplyr::lag()           masks stats::lag()
```

```

## x purrr::lift()          masks caret::lift()
## x randomForest::margin() masks ggplot2::margin()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
library(tree)

## Warning: package 'tree' was built under R version 4.4.1
library(gbm)

## Warning: package 'gbm' was built under R version 4.4.1
## Loaded gbm 2.2.2
## This version of gbm is no longer under development. Consider transitioning to gbm3, https://github.com/gbm-dev/gbm3
library(rpart)
library(rpart.plot)

## Warning: package 'rpart.plot' was built under R version 4.4.1
library(ggplot2)
library(MLmetrics)

## Warning: package 'MLmetrics' was built under R version 4.4.1
##
## Attaching package: 'MLmetrics'
##
## The following objects are masked from 'package:caret':
##
##   MAE, RMSE
##
## The following object is masked from 'package:base':
##
##   Recall
library(MLeval)

## Warning: package 'MLeval' was built under R version 4.4.1
customers <- read.csv("C:/Users/argon/Documents/Desktop Prime/MS Business Analytics/Summer Semester/STA
colnames(customers)

## [1] "Age"          "Gender"       "AnnualIncome"
## [4] "NumberOfPurchases" "ProductCategory" "TimeSpentOnWebsite"
## [7] "LoyaltyProgram" "DiscountsAvailed" "PurchaseStatus"
glimpse(customers)

## Rows: 1,500
## Columns: 9
## $ Age          <int> 40, 20, 27, 24, 31, 66, 39, 64, 43, 20, 66, 70, 54, ~
## $ Gender       <int> 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, ~
## $ AnnualIncome <dbl> 66120.27, 23579.77, 127821.31, 137798.62, 99300.96, ~
## $ NumberOfPurchases <int> 8, 4, 11, 19, 19, 14, 16, 13, 20, 16, 11, 11, 9, 17~
## $ ProductCategory <int> 0, 2, 2, 3, 1, 4, 3, 2, 1, 0, 1, 2, 2, 0, 1, 4, 4, ~
## $ TimeSpentOnWebsite <dbl> 30.568601, 38.240097, 31.633212, 46.167059, 19.8235~
## $ LoyaltyProgram <int> 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, ~
## $ DiscountsAvailed <int> 5, 5, 0, 4, 0, 2, 4, 0, 3, 5, 5, 4, 5, 1, 4, 0, 0, ~
## $ PurchaseStatus <int> 1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, ~

```

```
unique_customers <- customers %>%
  distinct()
glimpse(unique_customers)
```

```
## Rows: 1,388
## Columns: 9
## $ Age          <int> 40, 20, 27, 24, 31, 66, 39, 64, 43, 20, 66, 70, 54, ~
## $ Gender       <int> 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, ~
## $ AnnualIncome <dbl> 66120.27, 23579.77, 127821.31, 137798.62, 99300.96, ~
## $ NumberOfPurchases <int> 8, 4, 11, 19, 19, 14, 16, 13, 20, 16, 11, 11, 9, 17~
## $ ProductCategory <int> 0, 2, 2, 3, 1, 4, 3, 2, 1, 0, 1, 2, 2, 0, 1, 4, 4, ~
## $ TimeSpentOnWebsite <dbl> 30.568601, 38.240097, 31.633212, 46.167059, 19.8235~
## $ LoyaltyProgram <int> 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, ~
## $ DiscountsAvailed <int> 5, 5, 0, 4, 0, 2, 4, 0, 3, 5, 5, 4, 5, 1, 4, 0, 0, ~
## $ PurchaseStatus <int> 1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, ~
```

```
summary(unique_customers)
```

```
##      Age          Gender      AnnualIncome      NumberOfPurchases
## Min.   :18.00   Min.   :0.0000   Min.    : 20002   Min.    : 0.00
## 1st Qu.:30.75   1st Qu.:0.0000   1st Qu.: 53767   1st Qu.: 6.00
## Median :44.00   Median :1.0000   Median : 84625   Median :11.00
## Mean   :43.94   Mean    :0.5014   Mean    : 84699   Mean    :10.55
## 3rd Qu.:57.00   3rd Qu.:1.0000   3rd Qu.:117188   3rd Qu.:15.00
## Max.   :70.00   Max.    :1.0000   Max.    :149785   Max.    :20.00
## ProductCategory TimeSpentOnWebsite LoyaltyProgram DiscountsAvailed
## Min.   :0.000   Min.    : 1.037   Min.    :0.0000   Min.    :0.00
## 1st Qu.:1.000   1st Qu.:16.380   1st Qu.:0.0000   1st Qu.:1.00
## Median :2.000   Median :31.213   Median :0.0000   Median :3.00
## Mean   :2.003   Mean    :30.748   Mean    :0.3336   Mean    :2.61
## 3rd Qu.:3.000   3rd Qu.:44.666   3rd Qu.:1.0000   3rd Qu.:4.00
## Max.   :4.000   Max.    :59.991   Max.    :1.0000   Max.    :5.00
## PurchaseStatus
## Min.   :0.0000
## 1st Qu.:0.0000
## Median :0.0000
## Mean   :0.4669
## 3rd Qu.:1.0000
## Max.   :1.0000
```

```
BreakPointsDiscounts <- c(2)
BreakPointsTimeSpent <- c(-Inf, 29, Inf)
breakpoints <- c(-Inf, 40, Inf)
```

```
unique_customers <- unique_customers%>%
  mutate(
    Gender = as.factor(unique_customers$Gender),
    ProductCategory = as.factor(unique_customers$ProductCategory),
    LoyaltyProgram = as.factor(unique_customers$LoyaltyProgram),
    PurchaseStatus = factor(unique_customers$PurchaseStatus, levels = c(1, 0), labels = c("Yes", "No")),
    AnnualIncome = log(unique_customers$AnnualIncome),
    LoyaltyDiscInteraction = as.integer(unique_customers$LoyaltyProgram)*unique_customers$DiscountsAvailed,
    CatDiscountsAvailed = cut(unique_customers$DiscountsAvailed,
                              breaks = BreakPointsDiscounts,
                              labels = as.factor(c('Low Discounts Available', 'High Discounts Available')))
```

)

```
glimpse(unique_customers)
```

```
## Rows: 1,388
## Columns: 11
## $ Age                <int> 40, 20, 27, 24, 31, 66, 39, 64, 43, 20, 66, 70,~
## $ Gender             <fct> 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1,~
## $ AnnualIncome       <dbl> 11.09923, 10.06814, 11.75839, 11.83355, 11.5059~
## $ NumberOfPurchases <int> 8, 4, 11, 19, 19, 14, 16, 13, 20, 16, 11, 11, 9~
## $ ProductCategory    <fct> 0, 2, 2, 3, 1, 4, 3, 2, 1, 0, 1, 2, 2, 0, 1, 4,~
## $ TimeSpentOnWebsite <dbl> 30.568601, 38.240097, 31.633212, 46.167059, 19.~
## $ LoyaltyProgram     <fct> 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0,~
## $ DiscountsAvailed   <int> 5, 5, 0, 4, 0, 2, 4, 0, 3, 5, 5, 4, 5, 1, 4, 0,~
## $ PurchaseStatus     <fct> Yes, No, Yes, Yes, Yes, No, Yes, No, No, Yes, Y~
## $ LoyaltyDiscInteraction <int> 0, 0, 0, 0, 0, 0, 4, 0, 0, 0, 5, 0, 5, 1, 4, 0,~
## $ CatDiscountsAvailed <fct> High Discounts Availed, High Discounts Availed,~
```

```
set.seed(702)
```

```
train_ix = createDataPartition(unique_customers$PurchaseStatus, p=0.8)
```

```
unique_customers_train <- unique_customers[train_ix$Resample1,]
```

```
unique_customers_test  <- unique_customers[-train_ix$Resample1,]
```

```
train_ix
```

```
## $Resample1
##      [1]      2      3      5      6      7      9     10     11     12     13     14     15     16     17
##    [15]     18     19     20     21     23     24     25     26     27     28     29     30     31     32
##    [29]     33     34     35     36     37     38     39     41     42     43     44     46     47     48
##    [43]     50     51     53     54     55     56     57     58     59     60     61     63     64     65
##    [57]     66     67     69     70     71     72     73     75     76     77     79     80     81     82
##    [71]     83     84     85     86     87     89     90     91     92     93     96     98     99    101
##    [85]    102    106    107    108    109    110    111    112    113    114    115    116    118    119
##    [99]    120    121    122    123    125    126    128    129    132    133    134    135    136    137
##   [113]    138    139    140    141    142    143    144    145    146    147    148    150    152    154
##   [127]    155    156    158    161    162    163    164    165    166    167    168    169    170    171
##   [141]    172    173    174    175    177    178    179    180    181    182    183    186    187    188
##   [155]    189    190    191    192    193    194    195    196    197    199    200    201    203    204
##   [169]    206    207    209    210    211    214    215    216    217    218    219    220    221    222
##   [183]    224    225    226    227    228    229    230    231    232    233    234    235    236    237
##   [197]    239    242    244    245    248    249    251    252    253    254    255    256    257    258
##   [211]    261    262    263    264    267    268    269    270    271    272    274    276    278    279
##   [225]    280    281    282    283    284    286    287    289    290    291    293    294    295    296
##   [239]    297    300    301    302    303    304    305    306    307    309    310    311    312    313
##   [253]    314    315    317    318    319    320    321    323    324    325    326    328    329    330
##   [267]    331    332    333    335    336    338    340    341    342    343    344    345    346    347
##   [281]    348    349    351    352    353    355    357    358    359    360    361    362    363    364
##   [295]    365    366    367    368    369    370    371    373    375    376    377    378    379    380
##   [309]    381    383    386    387    388    389    390    392    393    395    398    401    402    403
##   [323]    404    407    408    409    410    411    412    413    415    416    417    418    419    420
##   [337]    422    424    425    426    430    431    432    435    436    439    440    442    443    445
##   [351]    447    448    449    450    451    452    453    455    456    457    458    459    460    461
##   [365]    462    463    464    465    466    467    468    469    470    471    472    473    474    475
##   [379]    477    478    479    480    481    482    483    484    485    487    488    489    490    492
```

##	[393]	493	494	495	496	497	498	499	500	501	503	505	506	507	508
##	[407]	511	513	514	516	517	519	520	521	522	523	525	526	527	528
##	[421]	532	533	534	535	536	537	538	539	540	541	542	543	545	546
##	[435]	547	548	549	550	551	552	553	555	556	557	558	559	560	561
##	[449]	562	563	564	565	567	568	569	570	571	572	573	574	575	576
##	[463]	577	579	583	585	586	587	588	589	590	591	592	593	594	595
##	[477]	596	597	598	599	600	601	603	604	605	606	607	608	609	611
##	[491]	612	613	614	615	616	617	618	620	621	622	623	624	625	627
##	[505]	628	629	632	633	634	635	636	638	639	642	644	645	646	647
##	[519]	648	649	650	653	656	657	658	659	660	661	662	664	665	666
##	[533]	667	668	669	670	671	672	674	677	678	680	681	682	685	686
##	[547]	687	689	690	691	692	693	695	699	700	701	703	704	705	707
##	[561]	709	711	712	713	715	716	718	720	721	722	723	724	726	727
##	[575]	728	729	730	731	732	734	735	736	737	738	739	740	741	742
##	[589]	743	746	747	748	749	751	752	753	754	755	757	758	759	760
##	[603]	761	762	763	764	765	767	768	769	771	772	773	774	775	776
##	[617]	777	780	781	782	783	784	786	787	788	790	791	792	795	798
##	[631]	799	800	801	804	805	806	807	808	809	810	811	812	813	814
##	[645]	816	818	819	820	821	823	824	825	826	827	832	833	834	835
##	[659]	836	837	839	840	842	843	845	846	848	849	850	851	852	856
##	[673]	857	859	860	861	862	863	865	866	867	868	869	870	871	872
##	[687]	873	874	875	877	878	881	882	883	884	885	886	887	888	889
##	[701]	890	891	892	893	894	895	896	897	899	900	901	902	903	904
##	[715]	905	906	907	908	909	911	912	913	914	915	916	917	918	919
##	[729]	920	921	923	925	926	927	928	929	930	931	933	935	937	938
##	[743]	940	942	944	946	947	948	949	950	953	954	955	956	957	958
##	[757]	959	960	961	962	963	965	966	967	968	969	970	971	972	973
##	[771]	974	976	977	978	979	980	982	983	984	985	986	987	988	990
##	[785]	991	992	993	994	995	996	997	998	999	1000	1001	1002	1003	1004
##	[799]	1005	1006	1007	1008	1009	1010	1011	1012	1013	1015	1016	1017	1018	1019
##	[813]	1020	1021	1022	1023	1024	1025	1026	1027	1029	1030	1031	1032	1033	1034
##	[827]	1035	1036	1037	1038	1039	1040	1041	1042	1043	1044	1045	1047	1048	1049
##	[841]	1052	1053	1054	1056	1057	1058	1059	1062	1065	1066	1067	1068	1069	1071
##	[855]	1072	1073	1074	1075	1076	1077	1079	1080	1082	1083	1084	1085	1086	1087
##	[869]	1088	1091	1092	1093	1094	1095	1096	1097	1098	1099	1101	1103	1104	1105
##	[883]	1106	1108	1109	1110	1112	1113	1114	1115	1117	1118	1119	1120	1121	1122
##	[897]	1123	1124	1125	1126	1128	1129	1130	1132	1133	1134	1135	1137	1140	1141
##	[911]	1142	1143	1144	1145	1146	1147	1148	1149	1152	1153	1154	1155	1156	1158
##	[925]	1161	1162	1163	1164	1165	1168	1170	1172	1173	1174	1175	1176	1177	1178
##	[939]	1179	1180	1181	1182	1183	1184	1185	1186	1188	1189	1190	1191	1195	1196
##	[953]	1197	1198	1199	1200	1201	1203	1204	1205	1207	1208	1209	1210	1211	1212
##	[967]	1213	1214	1215	1216	1217	1218	1219	1220	1222	1223	1224	1226	1227	1228
##	[981]	1229	1230	1231	1232	1233	1234	1235	1236	1237	1238	1239	1242	1243	1244
##	[995]	1245	1246	1247	1249	1250	1251	1253	1254	1255	1256	1257	1258	1259	1260
##	[1009]	1262	1263	1264	1266	1267	1268	1269	1270	1273	1275	1277	1278	1279	1280
##	[1023]	1281	1282	1283	1284	1285	1286	1287	1289	1290	1291	1292	1293	1294	1295
##	[1037]	1296	1298	1299	1300	1301	1302	1305	1306	1308	1309	1310	1311	1312	1313
##	[1051]	1314	1316	1317	1318	1319	1321	1322	1324	1325	1326	1327	1328	1330	1331
##	[1065]	1334	1335	1336	1338	1339	1341	1343	1344	1345	1347	1348	1349	1350	1351
##	[1079]	1352	1353	1355	1356	1358	1359	1360	1361	1362	1363	1364	1365	1366	1367
##	[1093]	1368	1370	1371	1372	1373	1375	1376	1377	1378	1379	1380	1381	1382	1383
##	[1107]	1384	1385	1386	1387	1388									

```

set.seed(702)
kcv <- 10
cv_folds <- createFolds(unique_customers_train$PurchaseStatus,
                        k = kcv)

my_summary = function(data, lev = NULL, model = NULL){
  default = defaultSummary(data, lev, model)
  twoclass = twoClassSummary(data, lev, model)

  twoclass[3] = 1 - twoclass[3]
  names(twoclass) = c('AUC_ROC', 'TPR', 'FPR')
  logloss = mnLogLoss(data, lev, model)
  c(default, twoclass, logloss)
}

fit_control <- trainControl(
  method = 'cv',
  indexOut = cv_folds,

  classProbs = TRUE,
  savePredictions = TRUE,
  summaryFunction = my_summary,
  selectionFunction = 'oneSE'
)

set.seed(702)
gbm_grid <- expand.grid(
  interaction.depth= c(7, 10, 12),
  n.trees = c(100,150,200,300),
  shrinkage = c(0.15, 0.2, 0.25),
  n.minobsinnode = c(5,7,10)
)

gbmfit <- train(
  PurchaseStatus~.-ProductCategory, data = unique_customers_train,
  method = 'gbm',
  trControl = fit_control,
  tuneGrid = gbm_grid,
  metric = 'logLoss',
  verbose = FALSE
)

gbmfit

## Stochastic Gradient Boosting
##
## 1111 samples
## 10 predictor
## 2 classes: 'Yes', 'No'
##
## No pre-processing
## Resampling: Cross-Validated (10 fold)
## Summary of sample sizes: 999, 1000, 1000, 1000, 999, 1000, ...
## Resampling results across tuning parameters:

```

##	shrinkage	interaction.depth	n.minobsinnode	n.trees	Accuracy	Kappa
##	0.15	7	5	100	0.9684841	0.9366671
##	0.15	7	5	150	0.9856094	0.9711085
##	0.15	7	5	200	0.9919159	0.9837862
##	0.15	7	5	300	0.9918998	0.9837441
##	0.15	7	7	100	0.9639713	0.9276249
##	0.15	7	7	150	0.9801960	0.9602355
##	0.15	7	7	200	0.9900978	0.9801305
##	0.15	7	7	300	0.9910069	0.9819511
##	0.15	7	10	100	0.9639552	0.9275851
##	0.15	7	10	150	0.9756752	0.9511862
##	0.15	7	10	200	0.9882881	0.9765016
##	0.15	7	10	300	0.9918917	0.9837252
##	0.15	10	5	100	0.9846763	0.9692354
##	0.15	10	5	150	0.9927845	0.9855047
##	0.15	10	5	200	0.9927926	0.9855239
##	0.15	10	5	300	0.9927926	0.9855239
##	0.15	10	7	100	0.9810967	0.9620725
##	0.15	10	7	150	0.9909907	0.9819134
##	0.15	10	7	200	0.9937016	0.9873617
##	0.15	10	7	300	0.9937016	0.9873617
##	0.15	10	10	100	0.9810806	0.9620217
##	0.15	10	10	150	0.9945944	0.9891501
##	0.15	10	10	200	0.9918998	0.9837354
##	0.15	10	10	300	0.9928007	0.9855506
##	0.15	12	5	100	0.9918998	0.9837354
##	0.15	12	5	150	0.9936935	0.9873473
##	0.15	12	5	200	0.9936935	0.9873432
##	0.15	12	5	300	0.9936935	0.9873432
##	0.15	12	7	100	0.9900818	0.9801018
##	0.15	12	7	150	0.9936935	0.9873390
##	0.15	12	7	200	0.9936935	0.9873473
##	0.15	12	7	300	0.9928007	0.9855593
##	0.15	12	10	100	0.9882880	0.9764802
##	0.15	12	10	150	0.9909989	0.9819409
##	0.15	12	10	200	0.9918998	0.9837437
##	0.15	12	10	300	0.9927926	0.9855363
##	0.20	7	5	100	0.9774931	0.9548393
##	0.20	7	5	150	0.9900818	0.9800962
##	0.20	7	5	200	0.9936935	0.9873514
##	0.20	7	5	300	0.9928007	0.9855589
##	0.20	7	7	100	0.9792788	0.9583955
##	0.20	7	7	150	0.9918916	0.9837245
##	0.20	7	7	200	0.9928007	0.9855547
##	0.20	7	7	300	0.9928007	0.9855547
##	0.20	7	10	100	0.9702697	0.9402135
##	0.20	7	10	150	0.9873791	0.9746466
##	0.20	7	10	200	0.9918917	0.9837252
##	0.20	7	10	300	0.9918917	0.9837293
##	0.20	10	5	100	0.9891889	0.9782826
##	0.20	10	5	150	0.9918998	0.9837391
##	0.20	10	5	200	0.9936935	0.9873473
##	0.20	10	5	300	0.9928007	0.9855547

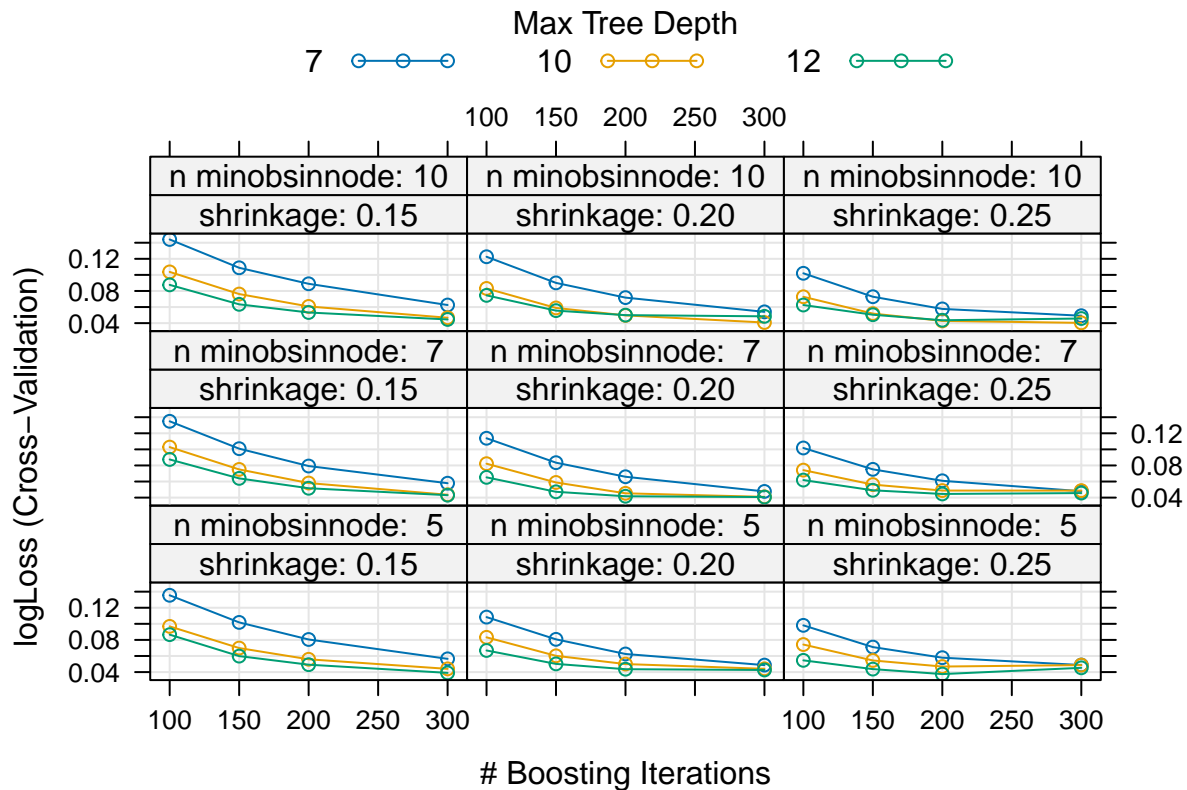
##	0.20	10	7	100	0.9909987	0.9819365
##	0.20	10	7	150	0.9936935	0.9873390
##	0.20	10	7	200	0.9936935	0.9873390
##	0.20	10	7	300	0.9936935	0.9873473
##	0.20	10	10	100	0.9873951	0.9747087
##	0.20	10	10	150	0.9918998	0.9837395
##	0.20	10	10	200	0.9937016	0.9873493
##	0.20	10	10	300	0.9945944	0.9891419
##	0.20	12	5	100	0.9928007	0.9855465
##	0.20	12	5	150	0.9937016	0.9873576
##	0.20	12	5	200	0.9928007	0.9855547
##	0.20	12	5	300	0.9928007	0.9855424
##	0.20	12	7	100	0.9918998	0.9837350
##	0.20	12	7	150	0.9927926	0.9855321
##	0.20	12	7	200	0.9927926	0.9855321
##	0.20	12	7	300	0.9927926	0.9855321
##	0.20	12	10	100	0.9892051	0.9783203
##	0.20	12	10	150	0.9874033	0.9747022
##	0.20	12	10	200	0.9882962	0.9764948
##	0.20	12	10	300	0.9892051	0.9783161
##	0.25	7	5	100	0.9856014	0.9711015
##	0.25	7	5	150	0.9909989	0.9819326
##	0.25	7	5	200	0.9901060	0.9801355
##	0.25	7	5	300	0.9901060	0.9801355
##	0.25	7	7	100	0.9855933	0.9710742
##	0.25	7	7	150	0.9909989	0.9819367
##	0.25	7	7	200	0.9928007	0.9855547
##	0.25	7	7	300	0.9928007	0.9855547
##	0.25	7	10	100	0.9819979	0.9638781
##	0.25	7	10	150	0.9918998	0.9837354
##	0.25	7	10	200	0.9918998	0.9837437
##	0.25	7	10	300	0.9937016	0.9873617
##	0.25	10	5	100	0.9936935	0.9873473
##	0.25	10	5	150	0.9918917	0.9837252
##	0.25	10	5	200	0.9936935	0.9873432
##	0.25	10	5	300	0.9928007	0.9855552
##	0.25	10	7	100	0.9918998	0.9837441
##	0.25	10	7	150	0.9919078	0.9837668
##	0.25	10	7	200	0.9919078	0.9837668
##	0.25	10	7	300	0.9910069	0.9819598
##	0.25	10	10	100	0.9918917	0.9837252
##	0.25	10	10	150	0.9928007	0.9855460
##	0.25	10	10	200	0.9937016	0.9873530
##	0.25	10	10	300	0.9918998	0.9837350
##	0.25	12	5	100	0.9927926	0.9855321
##	0.25	12	5	150	0.9936935	0.9873473
##	0.25	12	5	200	0.9954873	0.9909468
##	0.25	12	5	300	0.9936935	0.9873473
##	0.25	12	7	100	0.9909908	0.9819100
##	0.25	12	7	150	0.9918917	0.9837169
##	0.25	12	7	200	0.9936935	0.9873514
##	0.25	12	7	300	0.9918917	0.9837211
##	0.25	12	10	100	0.9918998	0.9837437
##	0.25	12	10	150	0.9900980	0.9801339

##	0.25	12	10	200	0.9918998	0.9837478
##	0.25	12	10	300	0.9909908	0.9819265
##	AUC_ROC	TPR	FPR	logLoss		
##	0.9931578	0.9653469	0.028700565	0.13550407		
##	0.9956384	0.9865008	0.015112994	0.10177174		
##	0.9963826	0.9980769	0.013446328	0.08055780		
##	0.9967744	0.9961538	0.011807910	0.05646836		
##	0.9927716	0.9633861	0.035480226	0.13491809		
##	0.9958292	0.9826546	0.021920904	0.10086337		
##	0.9965149	0.9922700	0.011779661	0.07926730		
##	0.9967117	0.9942308	0.011807910	0.05791833		
##	0.9903261	0.9633861	0.035480226	0.14406276		
##	0.9950489	0.9806938	0.028728814	0.10883694		
##	0.9951158	0.9923077	0.015225989	0.08893875		
##	0.9961899	0.9942308	0.010141243	0.06251973		
##	0.9958292	0.9865385	0.016920904	0.09682422		
##	0.9964165	0.9941931	0.008446328	0.06986486		
##	0.9964491	0.9942308	0.008446328	0.05591410		
##	0.9965143	0.9942308	0.008446328	0.04400189		
##	0.9958283	0.9884238	0.025282486	0.10280064		
##	0.9961864	0.9941931	0.011807910	0.07505148		
##	0.9961555	0.9980769	0.010112994	0.05810891		
##	0.9964810	0.9980769	0.010112994	0.04354182		
##	0.9949876	0.9845777	0.021949153	0.10371154		
##	0.9957059	0.9980769	0.008446328	0.07619878		
##	0.9958651	0.9942308	0.010112994	0.06075350		
##	0.9958005	0.9961538	0.010112994	0.04671386		
##	0.9960259	0.9942308	0.010112994	0.08656299		
##	0.9961563	0.9980769	0.010141243	0.05984330		
##	0.9961551	0.9961538	0.008446328	0.04920246		
##	0.9966114	0.9961538	0.008446328	0.03900934		
##	0.9950475	0.9961538	0.015225989	0.08746366		
##	0.9951464	0.9961538	0.008446328	0.06391157		
##	0.9955369	0.9980769	0.010141243	0.05162764		
##	0.9961581	0.9980769	0.011807910	0.04300183		
##	0.9959650	0.9903469	0.013502825	0.08759585		
##	0.9958998	0.9961538	0.013502825	0.06339739		
##	0.9958661	0.9961538	0.011807910	0.05323300		
##	0.9965148	0.9961538	0.010141243	0.04456562		
##	0.9949843	0.9826546	0.027005650	0.10836121		
##	0.9963857	0.9941931	0.013531073	0.08069818		
##	0.9962541	0.9980769	0.010141243	0.06250534		
##	0.9963186	0.9980769	0.011807910	0.04868015		
##	0.9944719	0.9807315	0.021949153	0.11383441		
##	0.9959642	0.9961161	0.011807910	0.08344272		
##	0.9961603	0.9980769	0.011807910	0.06593295		
##	0.9964503	0.9980769	0.011807910	0.04769340		
##	0.9928020	0.9633861	0.023615819	0.12253793		
##	0.9952424	0.9865008	0.011836158	0.09003823		
##	0.9956999	0.9961538	0.011836158	0.07174845		
##	0.9956017	0.9961538	0.011836158	0.05402108		
##	0.9951814	0.9865008	0.008446328	0.08322281		
##	0.9959926	0.9942308	0.010141243	0.06021954		
##	0.9959927	0.9980769	0.010141243	0.04997540		

##	0.9961550	0.9980769	0.011807910	0.04406573
##	0.9962881	0.9961161	0.013474576	0.08217529
##	0.9960910	0.9961538	0.008446328	0.05882392
##	0.9964828	0.9961538	0.008446328	0.04532449
##	0.9964152	0.9980769	0.010141243	0.04092821
##	0.9962209	0.9941931	0.018559322	0.08314068
##	0.9965479	0.9961538	0.011807910	0.05889590
##	0.9965473	0.9961538	0.008418079	0.04962793
##	0.9966440	0.9961538	0.006751412	0.04076310
##	0.9960887	0.9961538	0.010112994	0.06688664
##	0.9959588	0.9980769	0.010112994	0.05015860
##	0.9961876	0.9980769	0.011807910	0.04357248
##	0.9962842	0.9961538	0.010112994	0.04259100
##	0.9964574	0.9923077	0.008446328	0.06521359
##	0.9967460	0.9942308	0.008446328	0.04723201
##	0.9963885	0.9942308	0.008446328	0.04168658
##	0.9964520	0.9942308	0.008446328	0.04068586
##	0.9958296	0.9903846	0.011807910	0.07454577
##	0.9959274	0.9884615	0.013502825	0.05556379
##	0.9958949	0.9884615	0.011836158	0.04994822
##	0.9960916	0.9903846	0.011807910	0.04847732
##	0.9952161	0.9903469	0.018531073	0.09812685
##	0.9959625	0.9942308	0.011807910	0.07110021
##	0.9960911	0.9923077	0.011807910	0.05789867
##	0.9962867	0.9923077	0.011807910	0.04877690
##	0.9949483	0.9884238	0.016864407	0.10172616
##	0.9955086	0.9961538	0.013502825	0.07513621
##	0.9961871	0.9980769	0.011807910	0.06090906
##	0.9964152	0.9980769	0.011807910	0.04791231
##	0.9947992	0.9884615	0.023644068	0.10198085
##	0.9958764	0.9961538	0.011807910	0.07282870
##	0.9963578	0.9961538	0.011807910	0.05767433
##	0.9962854	0.9980769	0.010112994	0.04922581
##	0.9944375	0.9980769	0.010141243	0.07424330
##	0.9954763	0.9961538	0.011836158	0.05444465
##	0.9961243	0.9980769	0.010141243	0.04689647
##	0.9961922	0.9980769	0.011807910	0.04866062
##	0.9957346	0.9961538	0.011807910	0.07419453
##	0.9955352	0.9980769	0.013474576	0.05617536
##	0.9959589	0.9980769	0.013474576	0.04876241
##	0.9955671	0.9980769	0.015169492	0.04872267
##	0.9955677	0.9961538	0.011836158	0.07278096
##	0.9962528	0.9961538	0.010141243	0.05187597
##	0.9962535	0.9961538	0.008446328	0.04275905
##	0.9963519	0.9942308	0.010141243	0.04033936
##	0.9958969	0.9961538	0.010141243	0.05458979
##	0.9954393	0.9980769	0.010141243	0.04370399
##	0.9961238	0.9980769	0.006779661	0.03750566
##	0.9956668	0.9980769	0.010141243	0.04529068
##	0.9951773	0.9923077	0.010141243	0.06183388
##	0.9950482	0.9923077	0.008446328	0.04908034
##	0.9950789	0.9980769	0.010141243	0.04454717
##	0.9953077	0.9942308	0.010141243	0.04563646
##	0.9958028	0.9961538	0.011807910	0.06253578

```
## 0.9958989 0.9961538 0.015197740 0.05046673
## 0.9960925 0.9980769 0.013502825 0.04358287
## 0.9961896 0.9961538 0.013531073 0.04573337
##
## logLoss was used to select the optimal model using the one SE rule.
## The final values used for the model were n.trees = 150, interaction.depth =
## 12, shrinkage = 0.2 and n.minobsinnode = 7.
```

```
set.seed(702)
plot(gbmfit)
```



```
set.seed(702)
confusionMatrix(gbmfit)
```

```
## Cross-Validated (10 fold) Confusion Matrix
##
## (entries are percentual average cell counts across resamples)
##
##           Reference
## Prediction  Yes   No
##           Yes 46.4 0.5
##           No  0.3 52.8
##
## Accuracy (average) : 0.9928
```

```
set.seed(702)
thresholder(
  gbmfit,
```

```

threshold = 0.5,
final = TRUE,
statistics = c('Sensitivity',
               'Specificity')
)

```

```

##      n.trees interaction.depth shrinkage n.minobsinnode prob_threshold Sensitivity
## 1      150              12      0.2              7          0.5    0.9942308
##      Specificity
## 1      0.9915537

```

```

set.seed(702)
gbmfit_res = thresholder(gbmfit,
                        threshold = seq(0.0005, 1, by = 0.005),
                        final = TRUE)
gbmfit_res

```

```

##      n.trees interaction.depth shrinkage n.minobsinnode prob_threshold
## 1      150              12      0.2              7          0.0005
## 2      150              12      0.2              7          0.0055
## 3      150              12      0.2              7          0.0105
## 4      150              12      0.2              7          0.0155
## 5      150              12      0.2              7          0.0205
## 6      150              12      0.2              7          0.0255
## 7      150              12      0.2              7          0.0305
## 8      150              12      0.2              7          0.0355
## 9      150              12      0.2              7          0.0405
## 10     150              12      0.2              7          0.0455
## 11     150              12      0.2              7          0.0505
## 12     150              12      0.2              7          0.0555
## 13     150              12      0.2              7          0.0605
## 14     150              12      0.2              7          0.0655
## 15     150              12      0.2              7          0.0705
## 16     150              12      0.2              7          0.0755
## 17     150              12      0.2              7          0.0805
## 18     150              12      0.2              7          0.0855
## 19     150              12      0.2              7          0.0905
## 20     150              12      0.2              7          0.0955
## 21     150              12      0.2              7          0.1005
## 22     150              12      0.2              7          0.1055
## 23     150              12      0.2              7          0.1105
## 24     150              12      0.2              7          0.1155
## 25     150              12      0.2              7          0.1205
## 26     150              12      0.2              7          0.1255
## 27     150              12      0.2              7          0.1305
## 28     150              12      0.2              7          0.1355
## 29     150              12      0.2              7          0.1405
## 30     150              12      0.2              7          0.1455
## 31     150              12      0.2              7          0.1505
## 32     150              12      0.2              7          0.1555
## 33     150              12      0.2              7          0.1605
## 34     150              12      0.2              7          0.1655
## 35     150              12      0.2              7          0.1705
## 36     150              12      0.2              7          0.1755
## 37     150              12      0.2              7          0.1805

```

## 38	150	12	0.2	7	0.1855
## 39	150	12	0.2	7	0.1905
## 40	150	12	0.2	7	0.1955
## 41	150	12	0.2	7	0.2005
## 42	150	12	0.2	7	0.2055
## 43	150	12	0.2	7	0.2105
## 44	150	12	0.2	7	0.2155
## 45	150	12	0.2	7	0.2205
## 46	150	12	0.2	7	0.2255
## 47	150	12	0.2	7	0.2305
## 48	150	12	0.2	7	0.2355
## 49	150	12	0.2	7	0.2405
## 50	150	12	0.2	7	0.2455
## 51	150	12	0.2	7	0.2505
## 52	150	12	0.2	7	0.2555
## 53	150	12	0.2	7	0.2605
## 54	150	12	0.2	7	0.2655
## 55	150	12	0.2	7	0.2705
## 56	150	12	0.2	7	0.2755
## 57	150	12	0.2	7	0.2805
## 58	150	12	0.2	7	0.2855
## 59	150	12	0.2	7	0.2905
## 60	150	12	0.2	7	0.2955
## 61	150	12	0.2	7	0.3005
## 62	150	12	0.2	7	0.3055
## 63	150	12	0.2	7	0.3105
## 64	150	12	0.2	7	0.3155
## 65	150	12	0.2	7	0.3205
## 66	150	12	0.2	7	0.3255
## 67	150	12	0.2	7	0.3305
## 68	150	12	0.2	7	0.3355
## 69	150	12	0.2	7	0.3405
## 70	150	12	0.2	7	0.3455
## 71	150	12	0.2	7	0.3505
## 72	150	12	0.2	7	0.3555
## 73	150	12	0.2	7	0.3605
## 74	150	12	0.2	7	0.3655
## 75	150	12	0.2	7	0.3705
## 76	150	12	0.2	7	0.3755
## 77	150	12	0.2	7	0.3805
## 78	150	12	0.2	7	0.3855
## 79	150	12	0.2	7	0.3905
## 80	150	12	0.2	7	0.3955
## 81	150	12	0.2	7	0.4005
## 82	150	12	0.2	7	0.4055
## 83	150	12	0.2	7	0.4105
## 84	150	12	0.2	7	0.4155
## 85	150	12	0.2	7	0.4205
## 86	150	12	0.2	7	0.4255
## 87	150	12	0.2	7	0.4305
## 88	150	12	0.2	7	0.4355
## 89	150	12	0.2	7	0.4405
## 90	150	12	0.2	7	0.4455
## 91	150	12	0.2	7	0.4505

## 92	150	12	0.2	7	0.4555
## 93	150	12	0.2	7	0.4605
## 94	150	12	0.2	7	0.4655
## 95	150	12	0.2	7	0.4705
## 96	150	12	0.2	7	0.4755
## 97	150	12	0.2	7	0.4805
## 98	150	12	0.2	7	0.4855
## 99	150	12	0.2	7	0.4905
## 100	150	12	0.2	7	0.4955
## 101	150	12	0.2	7	0.5005
## 102	150	12	0.2	7	0.5055
## 103	150	12	0.2	7	0.5105
## 104	150	12	0.2	7	0.5155
## 105	150	12	0.2	7	0.5205
## 106	150	12	0.2	7	0.5255
## 107	150	12	0.2	7	0.5305
## 108	150	12	0.2	7	0.5355
## 109	150	12	0.2	7	0.5405
## 110	150	12	0.2	7	0.5455
## 111	150	12	0.2	7	0.5505
## 112	150	12	0.2	7	0.5555
## 113	150	12	0.2	7	0.5605
## 114	150	12	0.2	7	0.5655
## 115	150	12	0.2	7	0.5705
## 116	150	12	0.2	7	0.5755
## 117	150	12	0.2	7	0.5805
## 118	150	12	0.2	7	0.5855
## 119	150	12	0.2	7	0.5905
## 120	150	12	0.2	7	0.5955
## 121	150	12	0.2	7	0.6005
## 122	150	12	0.2	7	0.6055
## 123	150	12	0.2	7	0.6105
## 124	150	12	0.2	7	0.6155
## 125	150	12	0.2	7	0.6205
## 126	150	12	0.2	7	0.6255
## 127	150	12	0.2	7	0.6305
## 128	150	12	0.2	7	0.6355
## 129	150	12	0.2	7	0.6405
## 130	150	12	0.2	7	0.6455
## 131	150	12	0.2	7	0.6505
## 132	150	12	0.2	7	0.6555
## 133	150	12	0.2	7	0.6605
## 134	150	12	0.2	7	0.6655
## 135	150	12	0.2	7	0.6705
## 136	150	12	0.2	7	0.6755
## 137	150	12	0.2	7	0.6805
## 138	150	12	0.2	7	0.6855
## 139	150	12	0.2	7	0.6905
## 140	150	12	0.2	7	0.6955
## 141	150	12	0.2	7	0.7005
## 142	150	12	0.2	7	0.7055
## 143	150	12	0.2	7	0.7105
## 144	150	12	0.2	7	0.7155
## 145	150	12	0.2	7	0.7205

## 146	150	12	0.2	7	0.7255
## 147	150	12	0.2	7	0.7305
## 148	150	12	0.2	7	0.7355
## 149	150	12	0.2	7	0.7405
## 150	150	12	0.2	7	0.7455
## 151	150	12	0.2	7	0.7505
## 152	150	12	0.2	7	0.7555
## 153	150	12	0.2	7	0.7605
## 154	150	12	0.2	7	0.7655
## 155	150	12	0.2	7	0.7705
## 156	150	12	0.2	7	0.7755
## 157	150	12	0.2	7	0.7805
## 158	150	12	0.2	7	0.7855
## 159	150	12	0.2	7	0.7905
## 160	150	12	0.2	7	0.7955
## 161	150	12	0.2	7	0.8005
## 162	150	12	0.2	7	0.8055
## 163	150	12	0.2	7	0.8105
## 164	150	12	0.2	7	0.8155
## 165	150	12	0.2	7	0.8205
## 166	150	12	0.2	7	0.8255
## 167	150	12	0.2	7	0.8305
## 168	150	12	0.2	7	0.8355
## 169	150	12	0.2	7	0.8405
## 170	150	12	0.2	7	0.8455
## 171	150	12	0.2	7	0.8505
## 172	150	12	0.2	7	0.8555
## 173	150	12	0.2	7	0.8605
## 174	150	12	0.2	7	0.8655
## 175	150	12	0.2	7	0.8705
## 176	150	12	0.2	7	0.8755
## 177	150	12	0.2	7	0.8805
## 178	150	12	0.2	7	0.8855
## 179	150	12	0.2	7	0.8905
## 180	150	12	0.2	7	0.8955
## 181	150	12	0.2	7	0.9005
## 182	150	12	0.2	7	0.9055
## 183	150	12	0.2	7	0.9105
## 184	150	12	0.2	7	0.9155
## 185	150	12	0.2	7	0.9205
## 186	150	12	0.2	7	0.9255
## 187	150	12	0.2	7	0.9305
## 188	150	12	0.2	7	0.9355
## 189	150	12	0.2	7	0.9405
## 190	150	12	0.2	7	0.9455
## 191	150	12	0.2	7	0.9505
## 192	150	12	0.2	7	0.9555
## 193	150	12	0.2	7	0.9605
## 194	150	12	0.2	7	0.9655
## 195	150	12	0.2	7	0.9705
## 196	150	12	0.2	7	0.9755
## 197	150	12	0.2	7	0.9805
## 198	150	12	0.2	7	0.9855
## 199	150	12	0.2	7	0.9905

##	200	150	12	0.2	7	0.9955	
##		Sensitivity	Specificity	Pos Pred Value	Neg Pred Value	Precision	Recall
##	1	1.0000000	0.1233333	0.5004307	1.0000000	0.5004307	1.0000000
##	2	0.9980769	0.4140678	0.5994710	0.9966667	0.5994710	0.9980769
##	3	0.9980769	0.5373446	0.6549348	0.9971429	0.6549348	0.9980769
##	4	0.9980769	0.6420904	0.7104848	0.9975000	0.7104848	0.9980769
##	5	0.9980769	0.6927119	0.7407034	0.9975610	0.7407034	0.9980769
##	6	0.9980769	0.7400000	0.7716624	0.9978723	0.7716624	0.9980769
##	7	0.9980769	0.7821469	0.8011663	0.9980000	0.8011663	0.9980769
##	8	0.9980769	0.8124859	0.8238176	0.9980000	0.8238176	0.9980769
##	9	0.9980769	0.8327966	0.8398605	0.9980769	0.8398605	0.9980769
##	10	0.9980769	0.8462712	0.8508688	0.9980769	0.8508688	0.9980769
##	11	0.9980769	0.8699435	0.8709118	0.9981481	0.8709118	0.9980769
##	12	0.9980769	0.8868644	0.8858164	0.9981481	0.8858164	0.9980769
##	13	0.9980769	0.9003672	0.8979971	0.9981481	0.8979971	0.9980769
##	14	0.9980769	0.9020621	0.8995948	0.9981818	0.8995948	0.9980769
##	15	0.9980769	0.9122034	0.9091857	0.9981818	0.9091857	0.9980769
##	16	0.9980769	0.9172316	0.9139605	0.9981818	0.9139605	0.9980769
##	17	0.9980769	0.9273729	0.9237723	0.9982456	0.9237723	0.9980769
##	18	0.9980769	0.9324294	0.9286625	0.9982456	0.9286625	0.9980769
##	19	0.9980769	0.9341243	0.9303509	0.9982456	0.9303509	0.9980769
##	20	0.9980769	0.9357910	0.9321017	0.9982456	0.9321017	0.9980769
##	21	0.9980769	0.9374576	0.9337308	0.9982456	0.9337308	0.9980769
##	22	0.9980769	0.9425424	0.9386210	0.9982456	0.9386210	0.9980769
##	23	0.9980769	0.9442373	0.9403094	0.9982456	0.9403094	0.9980769
##	24	0.9980769	0.9510169	0.9472502	0.9982456	0.9472502	0.9980769
##	25	0.9980769	0.9527119	0.9490322	0.9982759	0.9490322	0.9980769
##	26	0.9980769	0.9577966	0.9544816	0.9983051	0.9544816	0.9980769
##	27	0.9980769	0.9577966	0.9544816	0.9983051	0.9544816	0.9980769
##	28	0.9980769	0.9594633	0.9561699	0.9983051	0.9561699	0.9980769
##	29	0.9980769	0.9611582	0.9579207	0.9983051	0.9579207	0.9980769
##	30	0.9980769	0.9628531	0.9595498	0.9983051	0.9595498	0.9980769
##	31	0.9980769	0.9645480	0.9612381	0.9983051	0.9612381	0.9980769
##	32	0.9980769	0.9679096	0.9648394	0.9983051	0.9648394	0.9980769
##	33	0.9980769	0.9679096	0.9648394	0.9983051	0.9648394	0.9980769
##	34	0.9980769	0.9679096	0.9648394	0.9983051	0.9648394	0.9980769
##	35	0.9980769	0.9696045	0.9666564	0.9983051	0.9666564	0.9980769
##	36	0.9980769	0.9696045	0.9666564	0.9983051	0.9666564	0.9980769
##	37	0.9980769	0.9712712	0.9684733	0.9983051	0.9684733	0.9980769
##	38	0.9980769	0.9712712	0.9684733	0.9983051	0.9684733	0.9980769
##	39	0.9980769	0.9712712	0.9684733	0.9983051	0.9684733	0.9980769
##	40	0.9980769	0.9729661	0.9702241	0.9983051	0.9702241	0.9980769
##	41	0.9980769	0.9729661	0.9702241	0.9983051	0.9702241	0.9980769
##	42	0.9980769	0.9729661	0.9702241	0.9983051	0.9702241	0.9980769
##	43	0.9980769	0.9729661	0.9702241	0.9983051	0.9702241	0.9980769
##	44	0.9980769	0.9746610	0.9720410	0.9983051	0.9720410	0.9980769
##	45	0.9980769	0.9763559	0.9738579	0.9983051	0.9738579	0.9980769
##	46	0.9980769	0.9763559	0.9738579	0.9983051	0.9738579	0.9980769
##	47	0.9980769	0.9780508	0.9757447	0.9983051	0.9757447	0.9980769
##	48	0.9980769	0.9780508	0.9757447	0.9983051	0.9757447	0.9980769
##	49	0.9980769	0.9780508	0.9757447	0.9983051	0.9757447	0.9980769
##	50	0.9980769	0.9780508	0.9757447	0.9983051	0.9757447	0.9980769
##	51	0.9980769	0.9780508	0.9757447	0.9983051	0.9757447	0.9980769
##	52	0.9961538	0.9780508	0.9756437	0.9965507	0.9756437	0.9961538

[illegible]

[illegible]

## 161	0.9884615	0.9949435	0.9942671	0.9899991	0.9942671	0.9884615
## 162	0.9884615	0.9949435	0.9942671	0.9899991	0.9942671	0.9884615
## 163	0.9884615	0.9949435	0.9942671	0.9899991	0.9942671	0.9884615
## 164	0.9865385	0.9949435	0.9942671	0.9883871	0.9942671	0.9865385
## 165	0.9865385	0.9949435	0.9942671	0.9883871	0.9942671	0.9865385
## 166	0.9865385	0.9949435	0.9942671	0.9883871	0.9942671	0.9865385
## 167	0.9865385	0.9949435	0.9942671	0.9883871	0.9942671	0.9865385
## 168	0.9865385	0.9949435	0.9942671	0.9883871	0.9942671	0.9865385
## 169	0.9846154	0.9949435	0.9942671	0.9868270	0.9942671	0.9846154
## 170	0.9826923	0.9949435	0.9942671	0.9852150	0.9942671	0.9826923
## 171	0.9807692	0.9949435	0.9942671	0.9835483	0.9942671	0.9807692
## 172	0.9807692	0.9949435	0.9942671	0.9835483	0.9942671	0.9807692
## 173	0.9788462	0.9949435	0.9942671	0.9818817	0.9942671	0.9788462
## 174	0.9750000	0.9949435	0.9942671	0.9786576	0.9942671	0.9750000
## 175	0.9711538	0.9949435	0.9942671	0.9754856	0.9942671	0.9711538
## 176	0.9673077	0.9949435	0.9942293	0.9722871	0.9942293	0.9673077
## 177	0.9673077	0.9949435	0.9942293	0.9722871	0.9942293	0.9673077
## 178	0.9634615	0.9949435	0.9942293	0.9692662	0.9942293	0.9634615
## 179	0.9576923	0.9949435	0.9942293	0.9645836	0.9942293	0.9576923
## 180	0.9500000	0.9949435	0.9941538	0.9583089	0.9941538	0.9500000
## 181	0.9423077	0.9949435	0.9940361	0.9521850	0.9940361	0.9423077
## 182	0.9365008	0.9949435	0.9939984	0.9475548	0.9939984	0.9365008
## 183	0.9249623	0.9949435	0.9939559	0.9387572	0.9939559	0.9249623
## 184	0.9230392	0.9949435	0.9939559	0.9373819	0.9939559	0.9230392
## 185	0.9095400	0.9949435	0.9939167	0.9275886	0.9939167	0.9095400
## 186	0.9037707	0.9949435	0.9939167	0.9236184	0.9939167	0.9037707
## 187	0.8941554	0.9949435	0.9939167	0.9166496	0.9939167	0.8941554
## 188	0.8903092	0.9949435	0.9939167	0.9137230	0.9939167	0.8903092
## 189	0.8845023	0.9966102	0.9958759	0.9095394	0.9958759	0.8845023
## 190	0.8671569	0.9966102	0.9957428	0.8972038	0.9957428	0.8671569
## 191	0.8402338	0.9966102	0.9956439	0.8790148	0.9956439	0.8402338
## 192	0.8228507	0.9966102	0.9955005	0.8665825	0.9955005	0.8228507
## 193	0.8016214	0.9966102	0.9953463	0.8526717	0.9953463	0.8016214
## 194	0.7765460	0.9966102	0.9951800	0.8371024	0.9951800	0.7765460
## 195	0.7496229	0.9966102	0.9951800	0.8214699	0.9951800	0.7496229
## 196	0.6955882	0.9983051	0.9972973	0.7906374	0.9972973	0.6955882
## 197	0.6243213	0.9983051	0.9970588	0.7536174	0.9970588	0.6243213
## 198	0.5568627	0.9983051	0.9966667	0.7216671	0.9966667	0.5568627
## 199	0.4797511	0.9983051	0.9960000	0.6874922	0.9960000	0.4797511
## 200	0.3428356	0.9983051	0.9933333	0.6347550	0.9933333	0.3428356
##	F1 Prevalence Detection Rate Detection Prevalence Balanced Accuracy					
## 1	0.6669056	0.4671487	0.4671487	0.9343223		0.5616667
## 2	0.7488400	0.4671487	0.4662478	0.7785115		0.7060724
## 3	0.7906158	0.4671487	0.4662478	0.7128089		0.7677108
## 4	0.8298177	0.4671487	0.4662478	0.6570004		0.8200837
## 5	0.8501616	0.4671487	0.4662478	0.6299968		0.8453944
## 6	0.8701512	0.4671487	0.4662478	0.6048034		0.8690385
## 7	0.8886683	0.4671487	0.4662478	0.5823453		0.8901119
## 8	0.9025074	0.4671487	0.4662478	0.5661773		0.9052814
## 9	0.9120511	0.4671487	0.4662478	0.5553500		0.9154368
## 10	0.9185113	0.4671487	0.4662478	0.5481669		0.9221741
## 11	0.9300586	0.4671487	0.4662478	0.5355540		0.9340102
## 12	0.9384870	0.4671487	0.4662478	0.5265530		0.9424707
## 13	0.9453308	0.4671487	0.4662478	0.5193373		0.9492221

## 14	0.9461973	0.4671487	0.4662478	0.5184364	0.9500695
## 15	0.9514518	0.4671487	0.4662478	0.5130309	0.9551402
## 16	0.9540700	0.4671487	0.4662478	0.5103443	0.9576543
## 17	0.9593710	0.4671487	0.4662478	0.5049387	0.9627249
## 18	0.9620218	0.4671487	0.4662478	0.5022441	0.9652532
## 19	0.9629218	0.4671487	0.4662478	0.5013432	0.9661006
## 20	0.9638387	0.4671487	0.4662478	0.5004503	0.9669339
## 21	0.9647222	0.4671487	0.4662478	0.4995574	0.9677673
## 22	0.9673730	0.4671487	0.4662478	0.4968547	0.9703096
## 23	0.9682730	0.4671487	0.4662478	0.4959538	0.9711571
## 24	0.9719238	0.4671487	0.4662478	0.4923502	0.9745469
## 25	0.9728402	0.4671487	0.4662478	0.4914493	0.9753944
## 26	0.9756428	0.4671487	0.4662478	0.4887384	0.9779368
## 27	0.9756428	0.4671487	0.4662478	0.4887384	0.9779368
## 28	0.9765427	0.4671487	0.4662478	0.4878456	0.9787701
## 29	0.9774597	0.4671487	0.4662478	0.4869447	0.9796176
## 30	0.9783431	0.4671487	0.4662478	0.4860438	0.9804650
## 31	0.9792431	0.4671487	0.4662478	0.4851429	0.9813125
## 32	0.9811122	0.4671487	0.4662478	0.4833409	0.9829933
## 33	0.9811122	0.4671487	0.4662478	0.4833409	0.9829933
## 34	0.9811122	0.4671487	0.4662478	0.4833409	0.9829933
## 35	0.9820466	0.4671487	0.4662478	0.4824400	0.9838407
## 36	0.9820466	0.4671487	0.4662478	0.4824400	0.9838407
## 37	0.9829811	0.4671487	0.4662478	0.4815472	0.9846741
## 38	0.9829811	0.4671487	0.4662478	0.4815472	0.9846741
## 39	0.9829811	0.4671487	0.4662478	0.4815472	0.9846741
## 40	0.9838980	0.4671487	0.4662478	0.4806463	0.9855215
## 41	0.9838980	0.4671487	0.4662478	0.4806463	0.9855215
## 42	0.9838980	0.4671487	0.4662478	0.4806463	0.9855215
## 43	0.9838980	0.4671487	0.4662478	0.4806463	0.9855215
## 44	0.9848324	0.4671487	0.4662478	0.4797454	0.9863690
## 45	0.9857668	0.4671487	0.4662478	0.4788445	0.9872164
## 46	0.9857668	0.4671487	0.4662478	0.4788445	0.9872164
## 47	0.9867192	0.4671487	0.4662478	0.4779436	0.9880639
## 48	0.9867192	0.4671487	0.4662478	0.4779436	0.9880639
## 49	0.9867192	0.4671487	0.4662478	0.4779436	0.9880639
## 50	0.9867192	0.4671487	0.4662478	0.4779436	0.9880639
## 51	0.9867192	0.4671487	0.4662478	0.4779436	0.9880639
## 52	0.9857494	0.4671487	0.4653469	0.4770427	0.9871023
## 53	0.9866658	0.4671487	0.4653469	0.4761418	0.9879498
## 54	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 55	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 56	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 57	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 58	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 59	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 60	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 61	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 62	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 63	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 64	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 65	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 66	0.9876182	0.4671487	0.4653469	0.4752409	0.9887973
## 67	0.9885704	0.4671487	0.4653469	0.4743400	0.9896447

##	68	0.9885704	0.4671487	0.4653469	0.4743400	0.9896447
##	69	0.9885704	0.4671487	0.4653469	0.4743400	0.9896447
##	70	0.9895048	0.4671487	0.4653469	0.4734471	0.9904781
##	71	0.9904389	0.4671487	0.4653469	0.4725462	0.9913255
##	72	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	73	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	74	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	75	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	76	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	77	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	78	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	79	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	80	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	81	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	82	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	83	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	84	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	85	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	86	0.9913912	0.4671487	0.4653469	0.4716534	0.9921588
##	87	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	88	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	89	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	90	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	91	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	92	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	93	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	94	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	95	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	96	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	97	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	98	0.9923436	0.4671487	0.4653469	0.4707525	0.9930063
##	99	0.9913733	0.4671487	0.4644460	0.4698516	0.9920448
##	100	0.9923257	0.4671487	0.4644460	0.4689507	0.9928922
##	101	0.9923257	0.4671487	0.4644460	0.4689507	0.9928922
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##	Accuracy	Kappa	J	Dist	
## 1	0.5328264	0.1163930	0.1233333	0.87666667	
## 2	0.6868354	0.3960236	0.4121447	0.58596856	
## 3	0.7525380	0.5187459	0.5354216	0.46269898	
## 4	0.8083465	0.6246167	0.6401673	0.35796411	
## 5	0.8353501	0.6763265	0.6907888	0.30734550	
## 6	0.8605435	0.7249754	0.7380769	0.26008376	
## 7	0.8830016	0.7685669	0.7802238	0.21796186	
## 8	0.8991696	0.8001131	0.8105628	0.18762287	
## 9	0.9099969	0.8213192	0.8308735	0.16733908	
## 10	0.9171800	0.8354267	0.8443481	0.15386451	
## 11	0.9297929	0.8602620	0.8680204	0.13023673	
## 12	0.9387939	0.8780445	0.8849413	0.11331583	
## 13	0.9460096	0.8923081	0.8984442	0.09981300	
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## 15	0.9523160	0.9048203	0.9102803	0.08801207	
## 16	0.9550026	0.9101509	0.9153086	0.08298382	
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## 21	0.9657895	0.9315982	0.9355346	0.06289388	
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## 23	0.9693931	0.9387646	0.9423142	0.05611422	
## 24	0.9729967	0.9459554	0.9490939	0.04933456	
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## 28	0.9775013	0.9549544	0.9575402	0.04140520	

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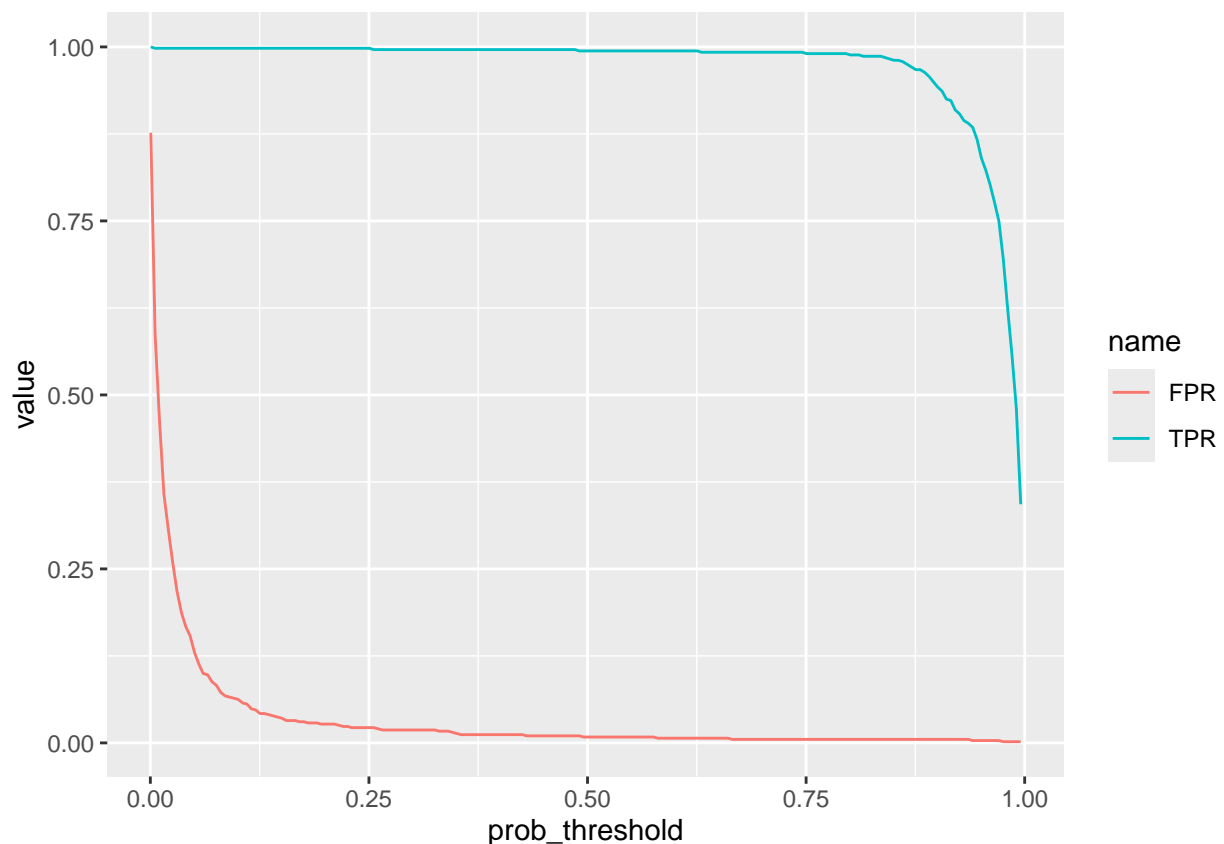
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## 190 0.9360916 0.8704200 0.8637670 0.13313047

```

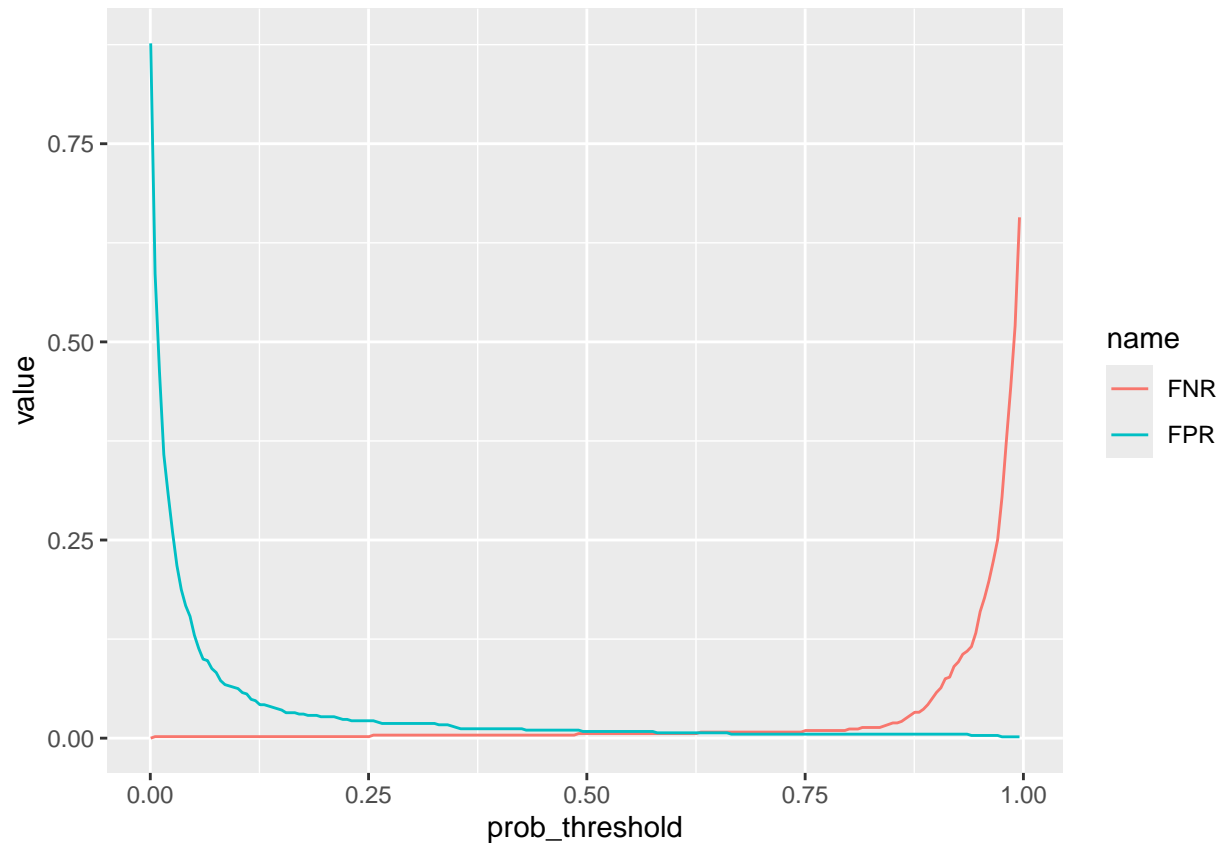
```
## 191 0.9235112 0.8446352 0.8368440 0.16003005
## 192 0.9153947 0.8280763 0.8194608 0.17734533
## 193 0.9054845 0.8077117 0.7982316 0.19853768
## 194 0.8937886 0.7835623 0.7731562 0.22358927
## 195 0.8811921 0.7574182 0.7462331 0.25051235
## 196 0.8568912 0.7068214 0.6938933 0.30446056
## 197 0.8235735 0.6369720 0.6226264 0.37571941
## 198 0.7920655 0.5702983 0.5551678 0.44317054
## 199 0.7560609 0.4935467 0.4780562 0.52027599
## 200 0.6921196 0.3552551 0.3411407 0.65718420
```

```
set.seed(702)
pldf = gbmfit_res %>%
  mutate(TPR = Sensitivity, FPR = 1 - Specificity, FNR = 1 - Sensitivity) %>%
  dplyr::select(-c(n.trees, interaction.depth, shrinkage, n.minobsinnode)) %>%
  pivot_longer(-prob_threshold)
```

```
set.seed(702)
ggplot(aes(x=prob_threshold, y=value, color = name),
  data = pldf %>% filter(name %in% c('TPR', 'FPR')) +
  geom_line()
```



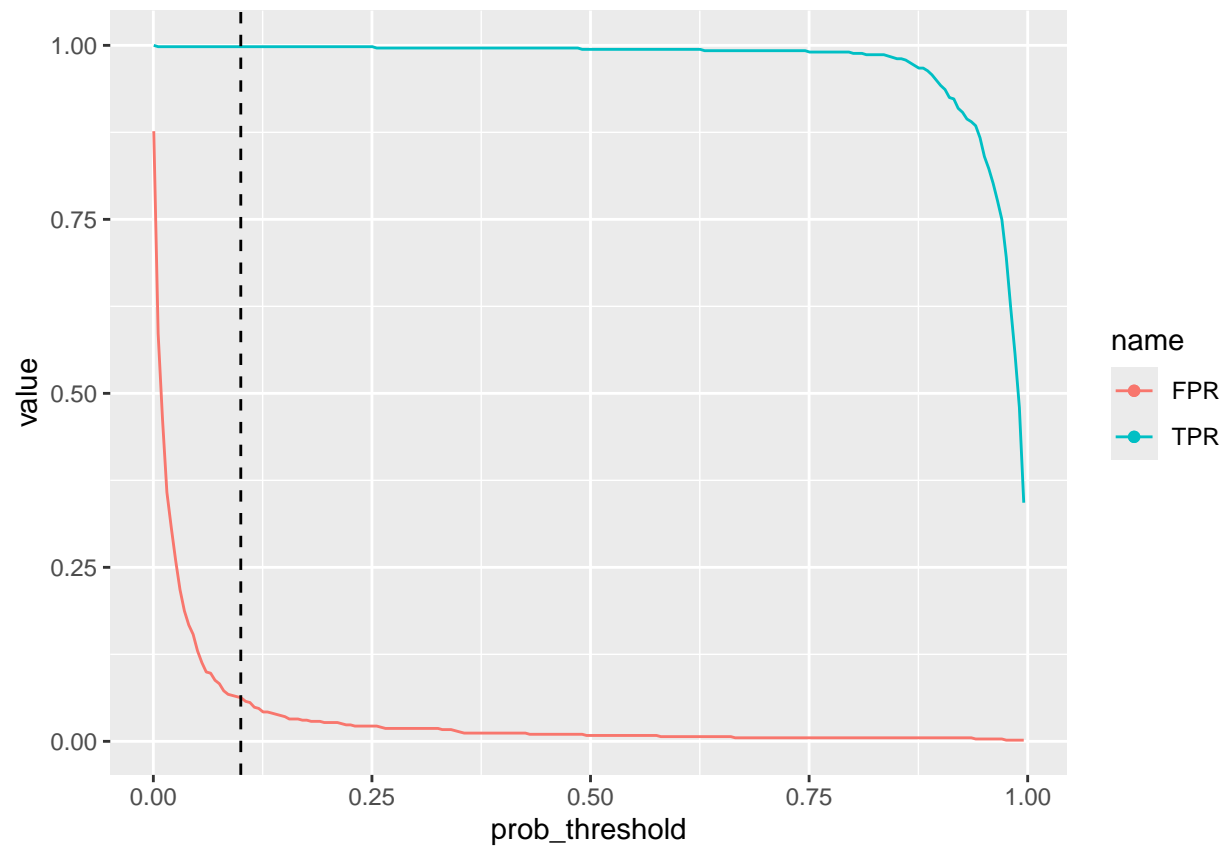
```
ggplot(aes(x=prob_threshold, y= value, color = name),
  data = pldf %>% filter(name %in% c('FNR', 'FPR')) +
  geom_line()
```



```
set.seed(702)
thres = 0.1

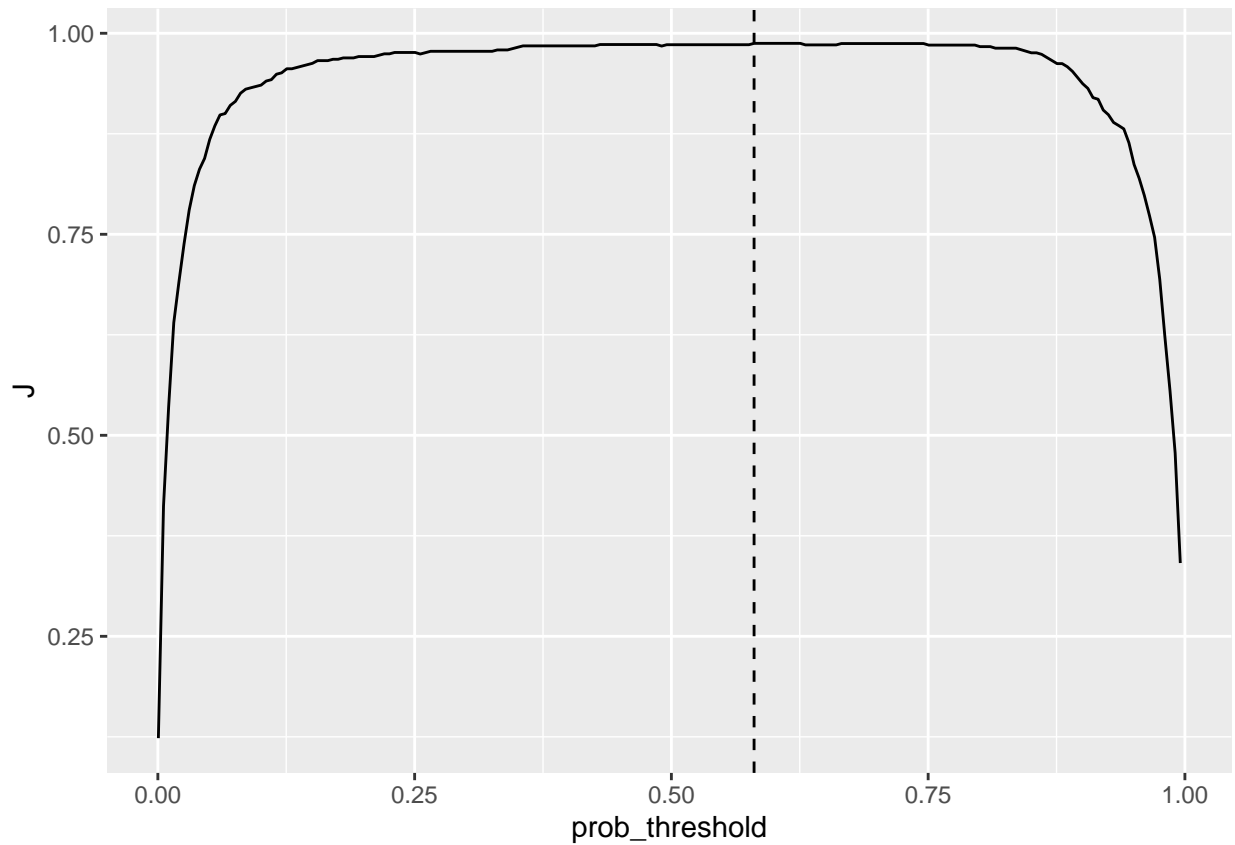
tp = gbmfit_res %>%
  dplyr::filter(prob_threshold==thres) %>%
  dplyr::select(prob_threshold, Sensitivity, Specificity) %>%
  mutate(TPR = Sensitivity, FPR = 1-Specificity)
```

```
set.seed(702)
ggplot(aes(x=prob_threshold, y=value, color = name),
  data = pldf %>% filter(name %in% c('TPR', 'FPR')) +
  geom_line() +
  geom_vline(xintercept = thres, lty = 2) +
  geom_point(aes(x=prob_threshold, y=TPR, color = NULL), data = tp) +
  geom_point(aes(x=prob_threshold, y = FPR, color = NULL), data = tp)
```

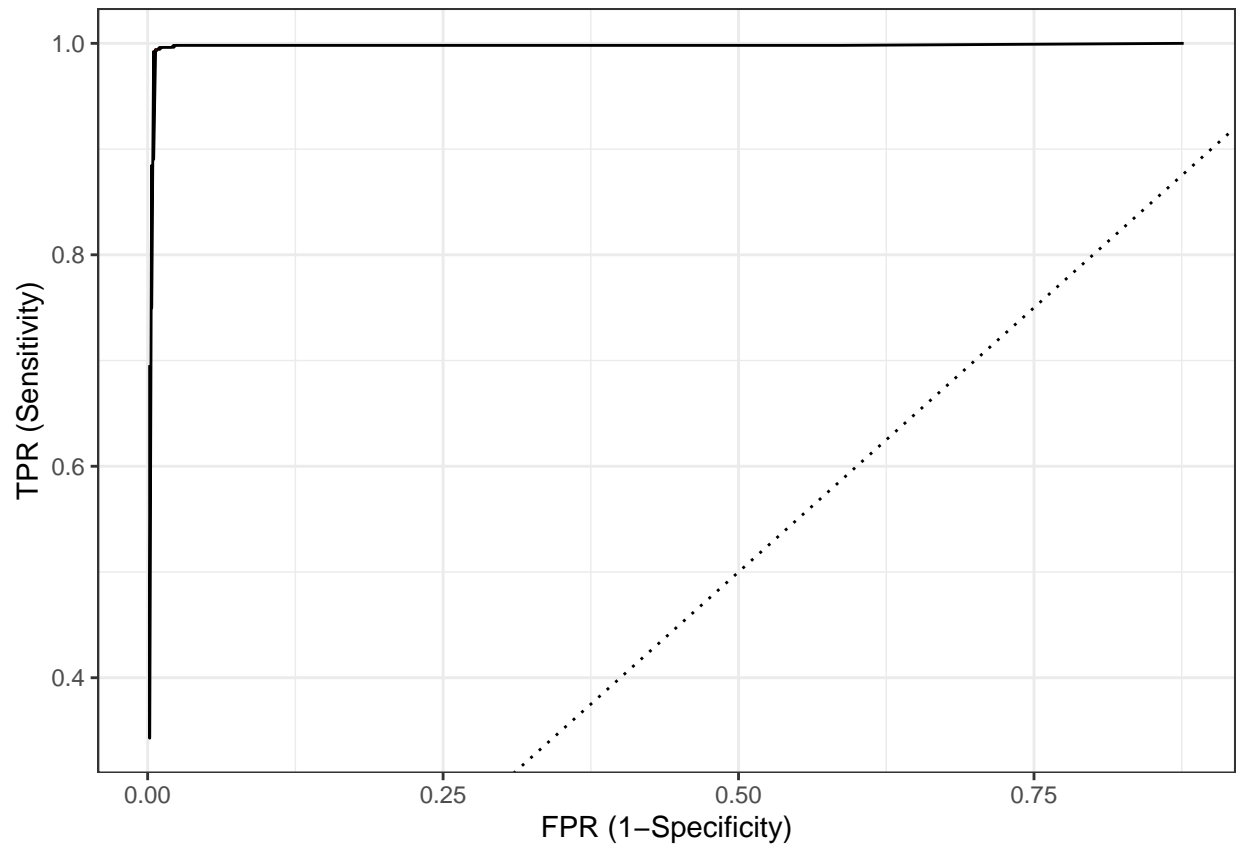


```
set.seed(702)
optim_J = gbmfit_res[which.max(gbmfit_res$J),]

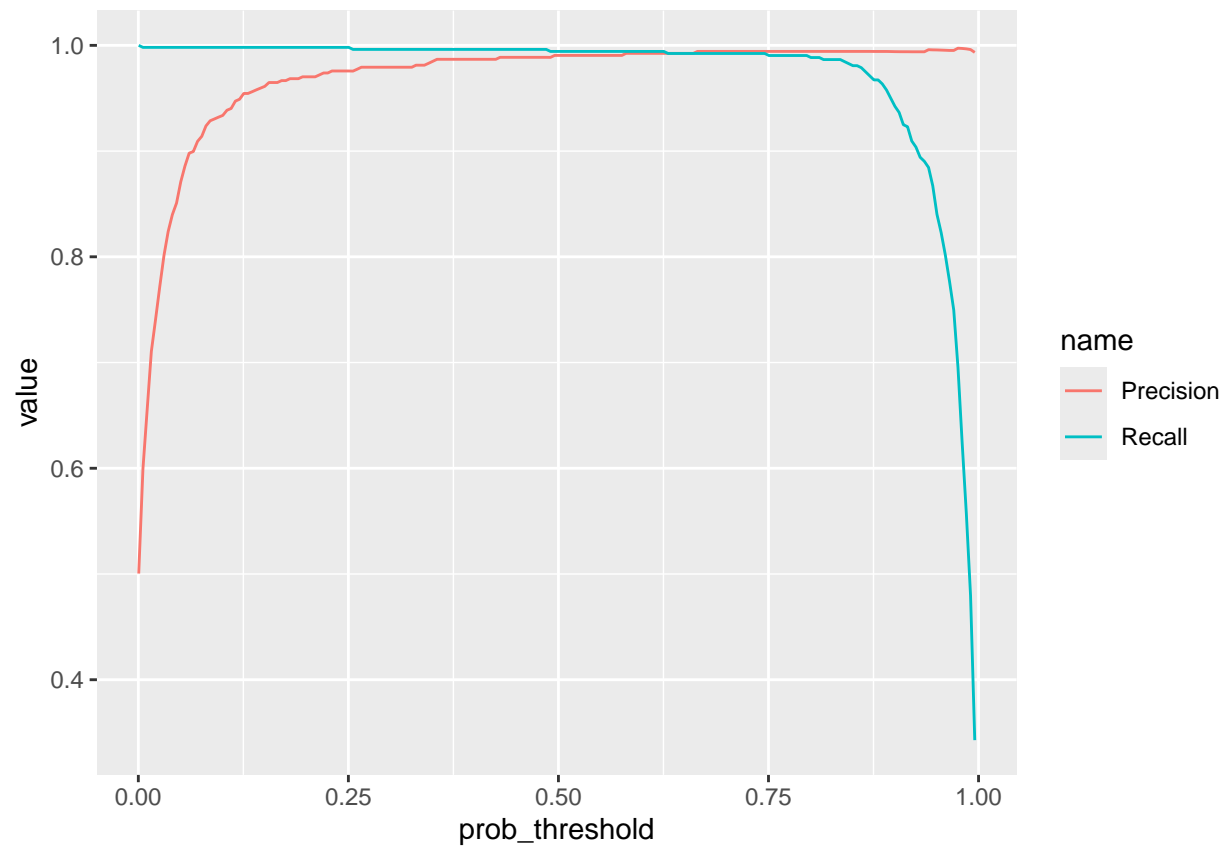
ggplot(aes(x=prob_threshold, y=J),
  data = gbmfit_res) +
  geom_line() +
  geom_vline(aes(xintercept = optim_J$prob_threshold), lty = 2)
```



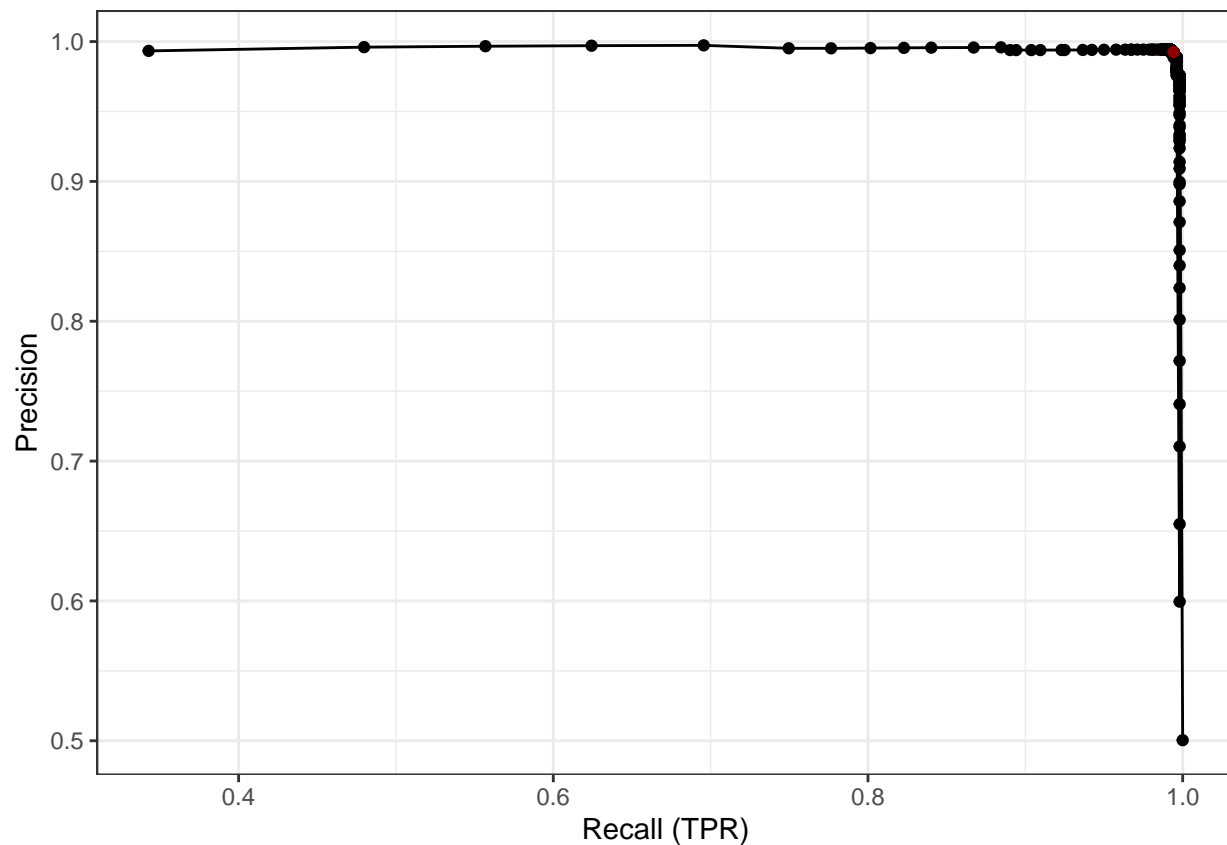
```
ggplot(aes(x=1 - Specificity, y = Sensitivity), data = gbmfit_res) +
  geom_line() +
  ylab('TPR (Sensitivity)') +
  xlab('FPR (1-Specificity)') +
  geom_abline(intercept = 0, slope = 1, linetype = 'dotted') +
  geom_segment(aes(x = 1-Specificity, xend=1-Specificity, y = Sensitivity, yend = Sensitivity), color =
```



```
set.seed(702)
ggplot(aes(x=prob_threshold, y=value, color = name),
  data=pldf %>% filter(name %in% c('Precision', 'Recall')))+
  geom_line()
```



```
ggplot(aes(x=Recall, y = Precision), data = gbmfit_res) +
  geom_point() +
  geom_line() +
  ylab('Precision') +
  xlab('Recall (TPR)') +
  geom_point(aes(x=Recall, y=Precision), color = 'darkred', data = optim_J) +
  theme_bw()
```

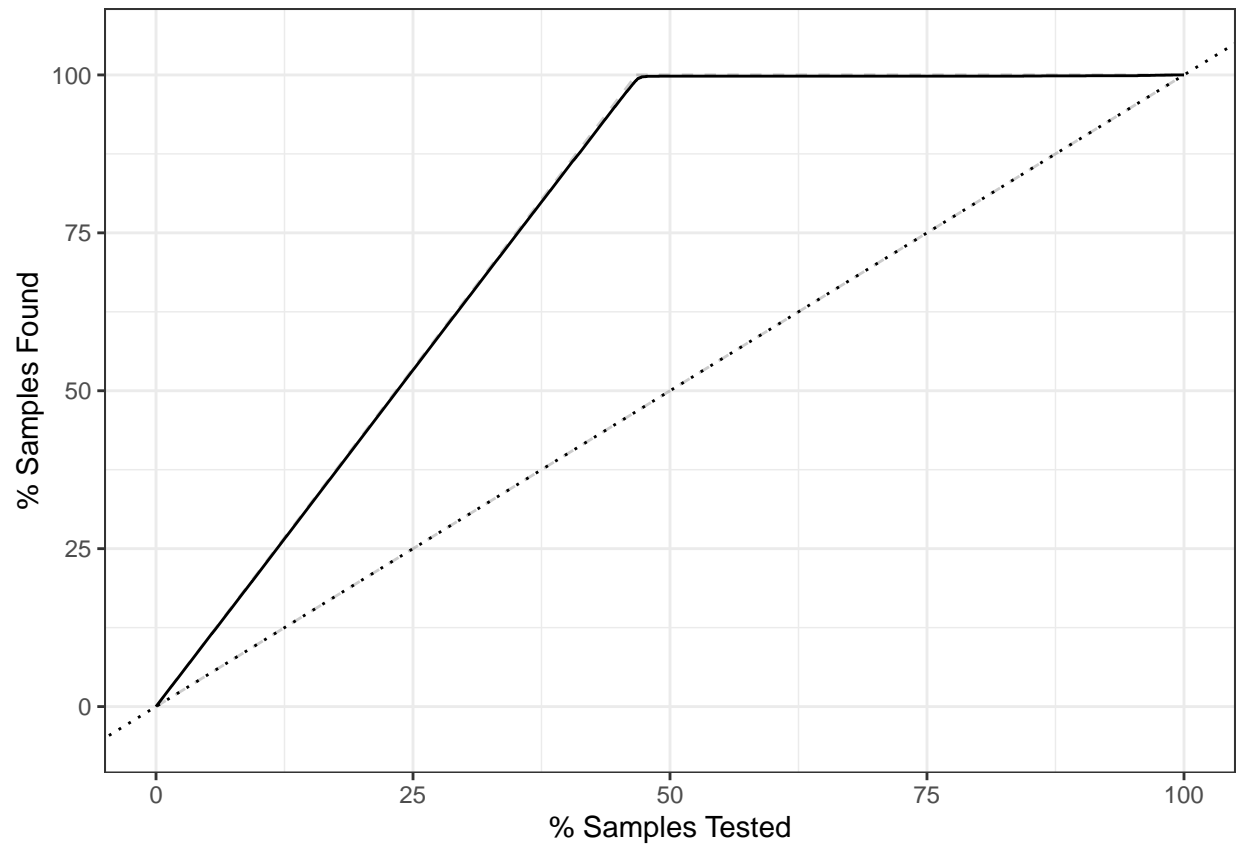
```
set.seed(702)
best_pars = gbmfit$bestTune
best_preds = gbmfit$pred %>% filter(n.trees==best_pars$n.trees, interaction.depth==best_pars$interaction.depth)

gbm_lift = caret::lift(obs~Yes, data = best_preds)
```

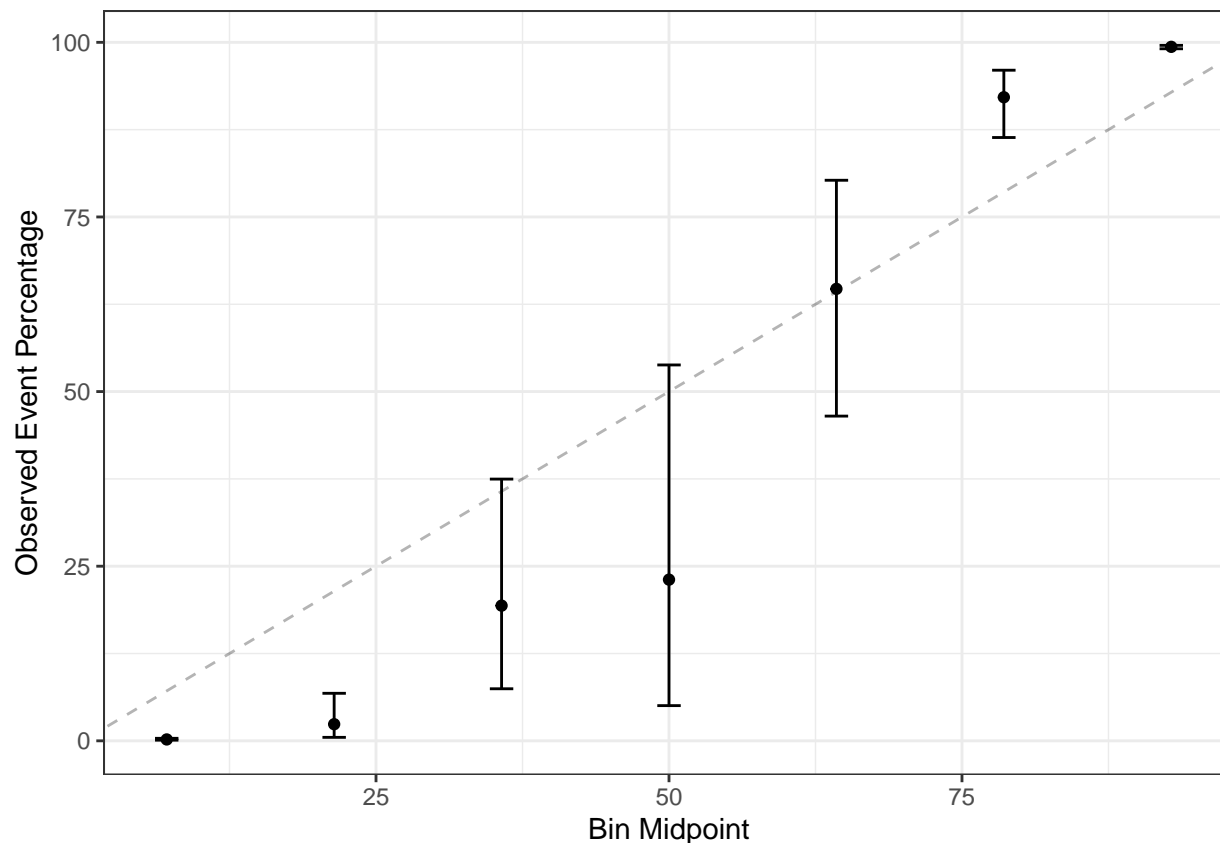
```
set.seed(702)
ggplot(gbm_lift) +
  geom_abline(slope = 1, linetype = "dotted") +
  xlim(c(0,100)) +
  theme_bw()
```

```
## Scale for x is already present.
```

```
## Adding another scale for x, which will replace the existing scale.
```



```
set.seed(702)
gbm_cal = caret::calibration(obs~Yes, data = best_preds, cuts = 7)
ggplot(gbm_cal) + theme_bw()
```



```
set.seed(702)
test_probs = predict(gbmfit, newdata = unique_customers_test, type='prob')

test_preds <- factor(ifelse(test_probs[, "Yes"] > optim_J$prob_threshold, "Yes", "No"))
test_conf_matrix = predict(gbmfit, newdata = unique_customers_test)

test_conf_matrix
```

```
##      [1] Yes Yes No Yes No Yes Yes No No No No Yes No No Yes Yes Yes Yes
##     [19] No Yes Yes Yes No Yes No No No No No Yes No Yes No Yes No No
##     [37] Yes No No Yes Yes Yes Yes No Yes Yes No Yes No Yes No No Yes Yes
##     [55] Yes Yes Yes Yes No No No No No No No No No Yes Yes No Yes Yes No
##     [73] No No Yes Yes No No No No No No No No No Yes Yes Yes Yes Yes
##     [91] No Yes Yes No No Yes Yes No No No No Yes No No No Yes No Yes Yes
##    [109] Yes No Yes Yes Yes Yes No No Yes No Yes No Yes Yes No No No Yes
##    [127] Yes No Yes No No No No No No No Yes Yes Yes No No No No Yes No
##    [145] Yes Yes Yes No No No No No No No No No No No No Yes No No No
##    [163] Yes No Yes Yes Yes Yes Yes No Yes Yes No No No No Yes Yes Yes Yes
##    [181] Yes No Yes No No Yes Yes No No Yes No No Yes No No No Yes Yes
##    [199] Yes No Yes Yes Yes Yes Yes Yes Yes No Yes No Yes Yes Yes No Yes No
##    [217] Yes Yes No No No Yes Yes No Yes No Yes Yes No No No No Yes Yes Yes
##    [235] No No Yes No No No No No Yes No Yes No No Yes Yes No Yes Yes No
##    [253] Yes No Yes No No Yes No No No No No No No Yes No Yes Yes Yes Yes
##    [271] No Yes No Yes Yes No Yes
## Levels: Yes No
```

```

print(confusionMatrix(test_preds, unique_customers_test$PurchaseStatus))

## Warning in confusionMatrix.default(test_preds,
## unique_customers_test$PurchaseStatus): Levels are not in the same order for
## reference and data. Refactoring data to match.

## Confusion Matrix and Statistics
##
##           Reference
## Prediction Yes  No
##           Yes 119  11
##           No   10 137
##
##           Accuracy : 0.9242
##           95% CI : (0.8864, 0.9525)
##           No Information Rate : 0.5343
##           P-Value [Acc > NIR] : <2e-16
##
##           Kappa : 0.8477
##
## Mcnemar's Test P-Value : 1
##
##           Sensitivity : 0.9225
##           Specificity : 0.9257
##           Pos Pred Value : 0.9154
##           Neg Pred Value : 0.9320
##           Prevalence : 0.4657
##           Detection Rate : 0.4296
##           Detection Prevalence : 0.4693
##           Balanced Accuracy : 0.9241
##
##           'Positive' Class : Yes
##
get_metrics = function(threshold, test_probs, true_class,
                        pos_label, neg_label){
  pc = factor(ifelse(test_probs[pos_label]>threshold, pos_label, neg_label), levels=c(pos_label, neg_label))
  test_set = data.frame(obs= true_class, pred=pc, test_probs)
  my_summary(test_set, lev= c(pos_label, neg_label))
}

get_metrics(optim_J$prob_threshold, test_probs, unique_customers_test$PurchaseStatus, 'Yes', 'No')

## Accuracy      Kappa      AUC_ROC      TPR      FPR      logLoss
## 0.92418773 0.84773447 0.93698931 0.92248062 0.07432432 0.35117326

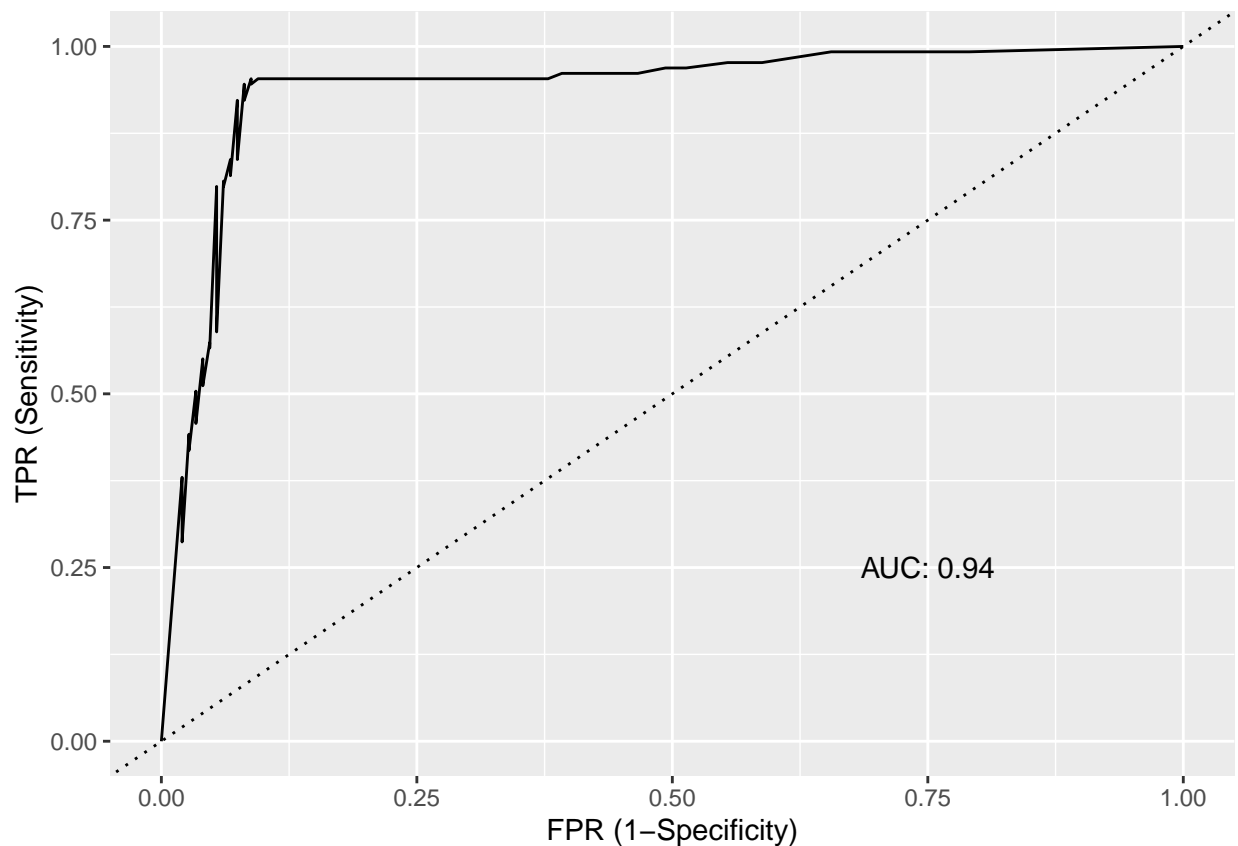
thr_seq = seq(0,1, length.out = 500)
metrics = lapply(thr_seq, function(x) get_metrics(x, test_probs, unique_customers_test$PurchaseStatus,
metrics_df = data.frame(do.call(rbind, metrics))
varImp(gbmfit)

## gbm variable importance
##
##
## Overall
## TimeSpentOnWebsite 100.00
## Age 68.77

```

```
## AnnualIncome                67.39
## NumberOfPurchases           53.66
## LoyaltyDiscInteraction       41.86
## LoyaltyProgram1             37.57
## DiscountsAvailed            31.13
## CatDiscountsAvailedHigh Discounts Availed  11.45
## Gender1                     0.00
```

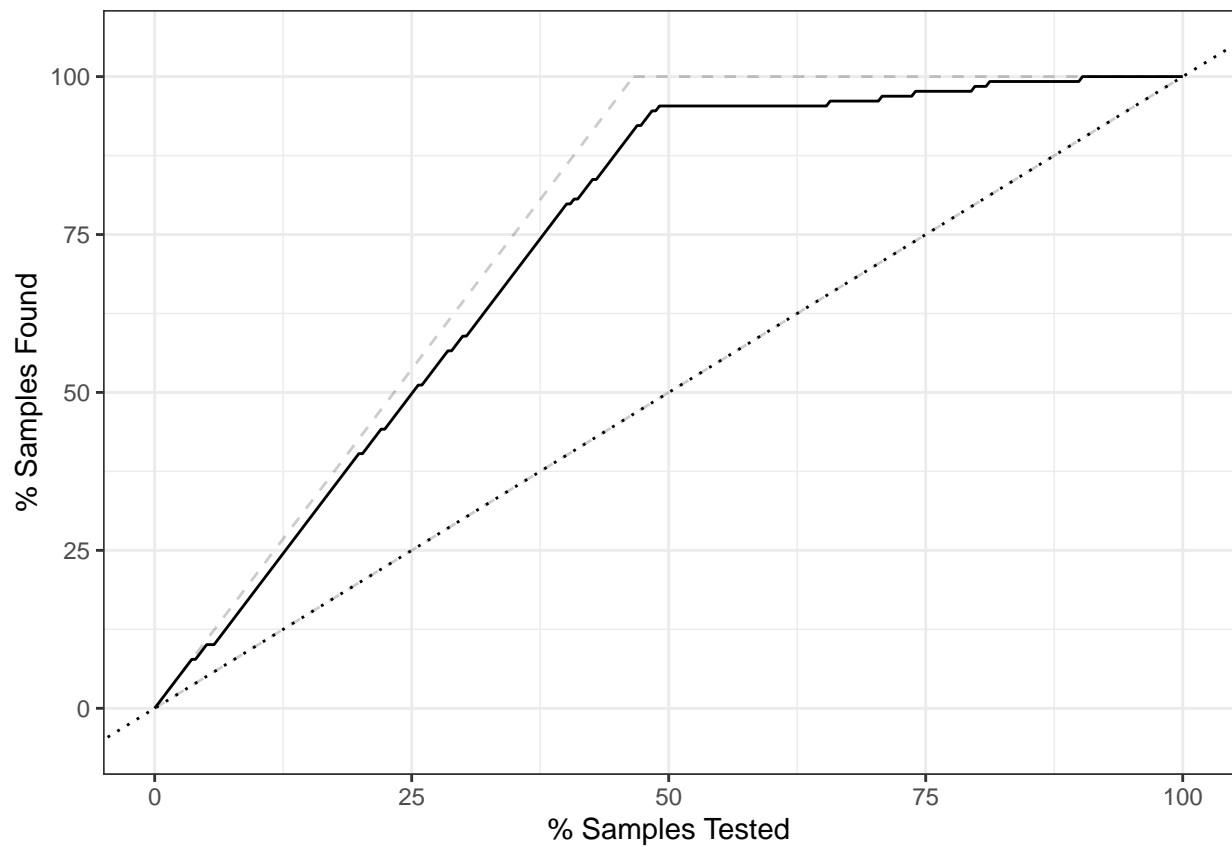
```
ggplot(aes(x=FPR, y = TPR), data = metrics_df) +
  geom_line() +
  ylab('TPR (Sensitivity)') +
  xlab('FPR (1-Specificity)') +
  geom_abline(intercept = 0, slope = 1, linetype = 'dotted') +
  annotate('text', x=0.75, y=0.25, label = paste('AUC:', round(metrics_df$AUC_ROC[1], 2)))
```



```
gbm_oos_lift = caret::lift(unique_customers_test$PurchaseStatus~test_probs[,1])

ggplot(gbm_oos_lift) +
  geom_abline(slope = 1, linetype = 'dotted') +
  xlim(c(0,100)) +
  theme_bw()
```

```
## Scale for x is already present.
## Adding another scale for x, which will replace the existing scale.
```



```
gbm_cal = caret::calibration(unique_customers_test$PurchaseStatus~test_probs[,1], data = best_preds, cu
ggplot(gbm_cal) + theme_bw()
```

