

# Machine learning Data Competition 2020

## Report I.

Shreyasvi Natraj

## 1 Introduction

For the given data competition, we were provided with data pertaining to previous advertisement campaigns as well as demographics of users who have been a part of the survey conducted.

For the objective of the given task, we are required to train a model that can be used in order to predict whether if a user is likely to have a “conversion” where a “conversion” refers to the user clicking on the advertisement and subscribing to the service.

Since the data provided is used in order to predict a categorical variable i.e. “conversion/y”, we planned to use do a quick an dirty implementation of the following models to check for their accuracy:

*K-Nearest Neighbours* Random Forest *LDA*, *QDA* & *C5.0* Supported Vector Machines \*Logistic Regression

We initially carried out with exploratory data analysis for the data provided to us by considering na values as NaN values.

## 2 Exploratory data analysis

We observed from this that it would not be a good idea to not consider the na values as NaN but as a separate level. However, based on similarity in between the classes, we can merge different levels of a factor into lesser number of levels so they are easier for our model to interpret.

### 2.1 Interpretation

We also carried out the same process of data analysis after converting categorical variables into integer format which tend to show a similar fashion to the current analysis being carried out. However, we replaced “na” values with 0 which made the data much more consistent. We also observed that `time_spent`, `outcome_old` and `X3` tends to hold a very high significance when predicting conversion `y`.

### 2.2 Plots

```
#install.packages("DataExplorer")
library(tidyverse)
library(DataExplorer)
dataset = read.csv('train.csv')
dataset[ dataset == "na" ] <- NA

web<- dataset
glimpse(web)
```

```
## Rows: 8,526
## Columns: 17
```

```
## $ age      <int> 35, 42, 38, 71, 37, 26, 27, 28, 57, 44, 34, 26, 33...
## $ job      <fct> manager, manager, industrial_worker, retired, unem...
## $ marital  <fct> single, married, married, married, single, married...
## $ education <fct> grad_school, grad_school, university, high_school,...
## $ device   <fct> NA, smartphone, NA, smartphone, desktop, smartphon...
## $ day      <int> 30, 17, 14, 13, 27, 17, 30, 7, 26, 28, 12, 2, 6, 2...
## $ month    <int> 5, 7, 5, 11, 4, 7, 6, 5, 5, 12, 8, 12, 5, 5, 11, 2...
## $ time_spent <dbl> 37.65, 39.25, 10.50, 8.80, 20.80, 57.00, 3.75, 10....
## $ banner_views <int> 1, 1, 2, 2, 2, 3, 5, 1, 5, 1, 1, 1, 1, 2, 1, 2, 1,...
## $ banner_views_old <int> 0, 0, 0, 1, 3, 0, 0, 1, 0, 0, 0, 0, 3, 0, 0, 0, 5,...
## $ days_elapsed_old <int> -1, -1, -1, 98, 179, -1, -1, 339, -1, -1, -1, -1, ...
## $ outcome_old <fct> NA, NA, NA, success, other, NA, NA, other, NA, NA,...
## $ X1          <int> 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,...
## $ X2          <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,...
## $ X3          <int> 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0,...
## $ X4          <dbl> 0.07793293, 0.07283969, 0.07607176, 0.09640840, 0....
## $ y          <int> 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1,...
```

*#To go with glimpse(), DataExplorer itself has got a function called introduce()*  
`introduce(web)`

```
##   rows columns discrete_columns continuous_columns all_missing_columns
## 1 8526      17           5              12              0
##   total_missing_values complete_rows total_observations memory_usage
## 1              8911          1917          144942          655976
```

*#The same introduce() could also be plotted in a pretty graph.*

```
plot_intro(web, ggtheme = theme_dark(),
            title = "EDA with Data Explorer",
            )
```

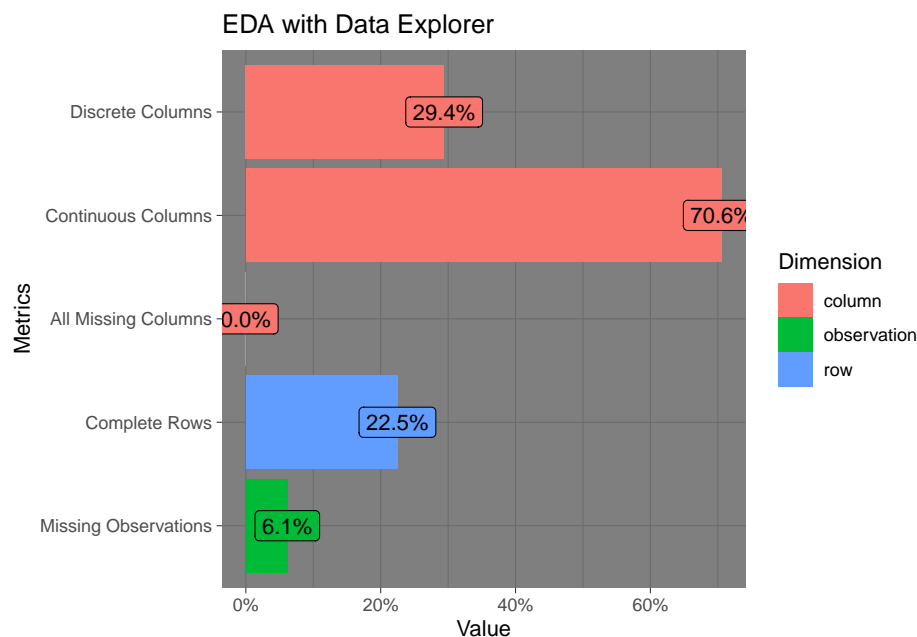


Figure 1: Data Distribution

```
plot_missing(web,
  ggtheme = theme_dark(),
  title = " Features missing from the whole observations",
)
```

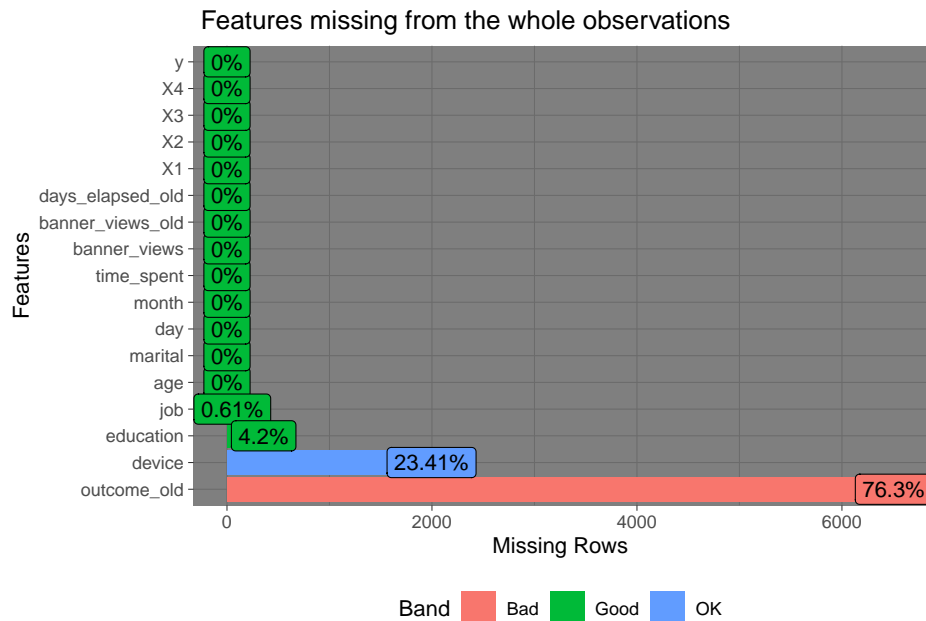


Figure 2: Missing Columns

###EDA for Continuous variables

*##for univariate*

```
DataExplorer::plot_histogram(web,
  ggtheme = theme_dark(),
  title = " Histogram of continuous features",
)
```

```
plot_density(web,
  ggtheme = theme_dark(),
  title = " Density of continuous features",
) # age, time_spent, X4 are right skewed
```

*#outcome old hase a mode which is na-> remove this category? or remove this variable??*

```
plot_bar(web,
  ggtheme = theme_dark(),
  title = " Density of continuous features",
) ##VISUALIZE DATA WHEN X2=0 AND =1 (subsetting)
```

```
a<- filter(web, web$X1==0)
b<- filter(web, web$X1==1)
plot_bar(a)
plot_bar(b)
plot_density(a,
```

```

        title = " a")
    plot_density(b,
        title = " b")

##for bivariate
    plot_boxplot(web, by= 'day' , ncol = 1,
        ggtheme = theme_dark(),
        title = " Boxplot of continuous features by day",
    )

####Correlation Plot
##autocorr plot
plot_correlation(web, cor_args = list( 'use' = 'complete.obs'),
    ggtheme = theme_dark(),
    title = " Autocorr Plot",
)

##continurous correlation plot
plot_correlation(web, type = 'c', cor_args = list( 'use' = 'complete.obs'),
    ggtheme = theme_dark(),
    title = " Continuous corr Plot",
)

####EDA for Categorical
plot_bar(a, maxcat = 390, parallel = FALSE,
    ggtheme = theme_dark(),
    title = " Categorical Features Plot",
)

```

For more sophisticated graphs, that span over multiple pages, see function `ggarrange()` from `ggpubr` package (see [link](#)).

For good-looking colors, have a look at the Paul Tol's palette <https://personal.sron.nl/~pault/>.

## 2.3 Tables

To display a table, look at the `kable()` function from `knitr` package. Also, consider the `kableExtra` package for more sophisticated options (see [link](#)). In Table 1, we show an example that uses both `kable` and `kableExtra`.

You can reference a table by putting the code `\\label{tab:tblname}` inside the caption. See code below. Then, you see that the reference works (see Table 1).

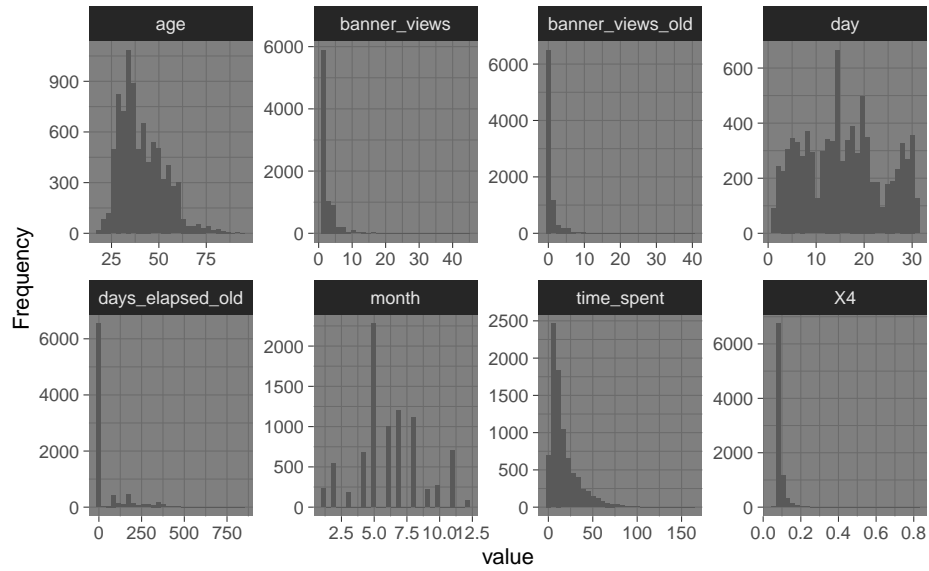
```

# Prepare data to put in the table
dat2 <- mtcars %>%
  group_by(cyl) %>%
  summarise(Average = mean(mpg), Max = max(mpg), Sqrt = sum(sqrt(mpg)))

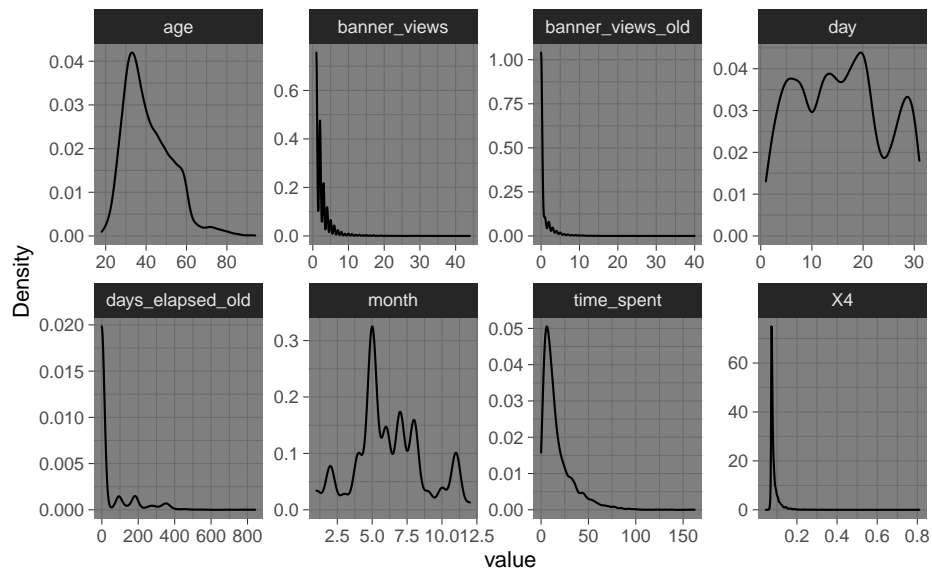
# Print table
capt <- paste("\\label{tab:tblname}Average and ",
    "maximum miles per gallon for each number of cylindyers class.")
kable(dat2,
    format = "latex",
    longtable = F,
    booktabs = T,

```

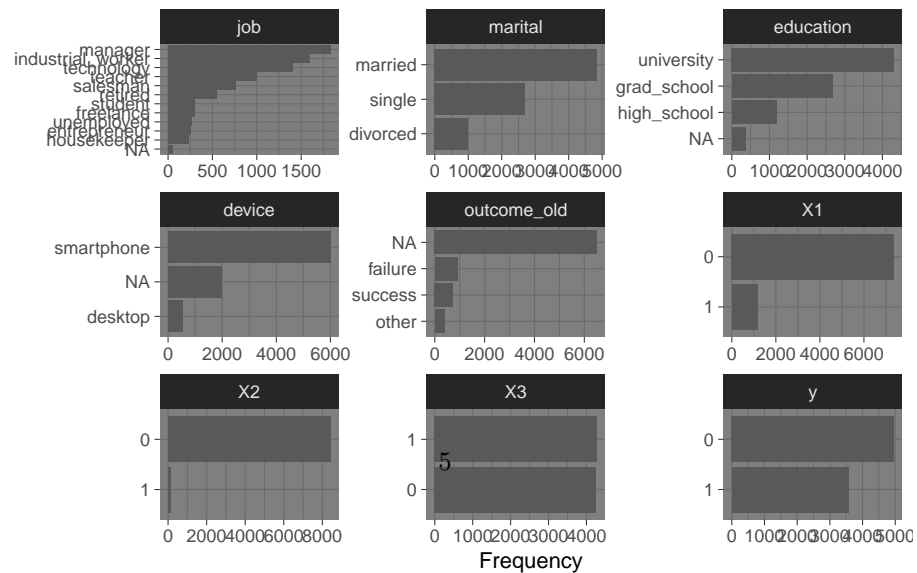
### Histogram of continuous features



### Density of continuous features



### Density of continuous features



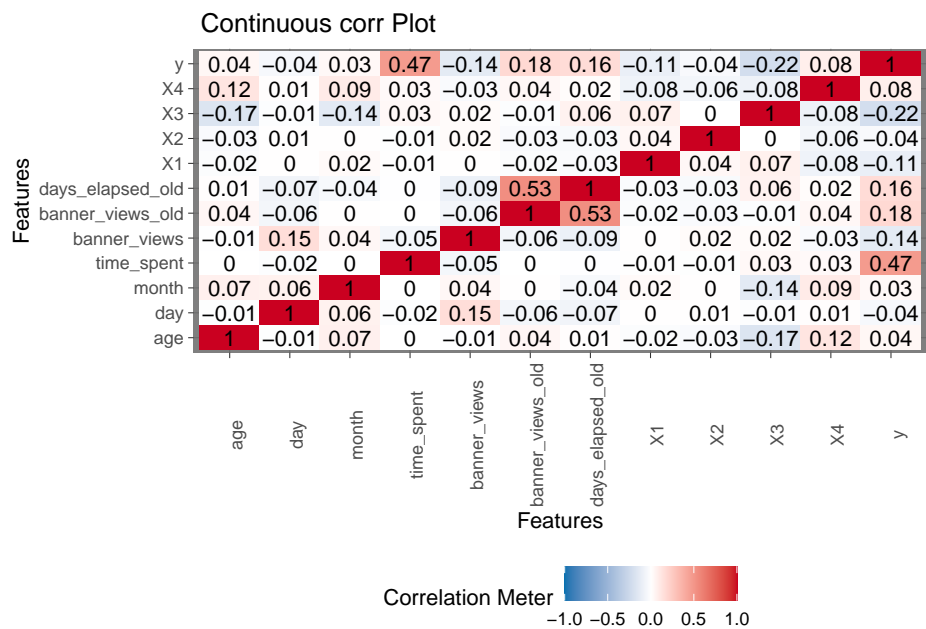
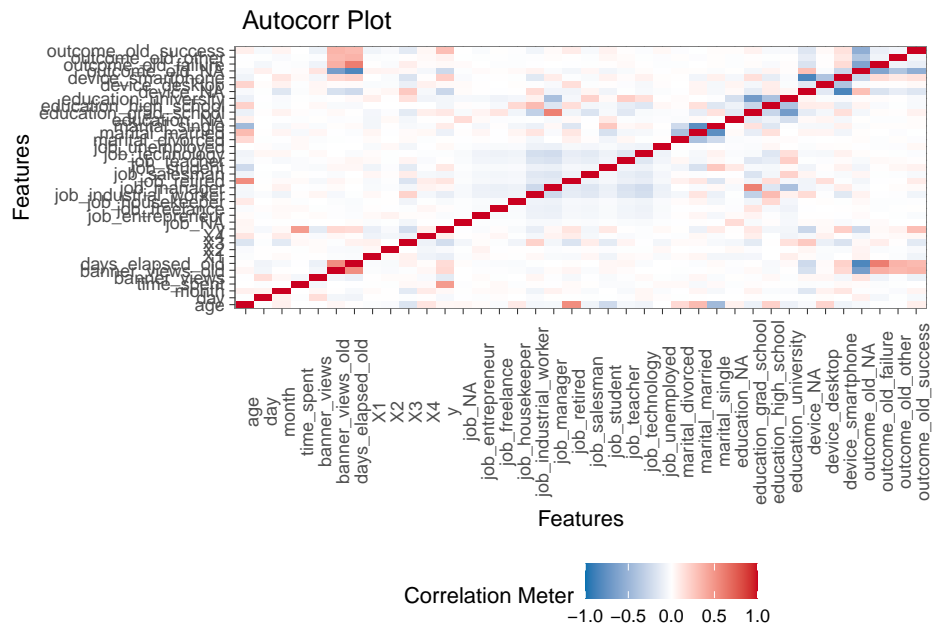


Figure 4: Correlation Plot

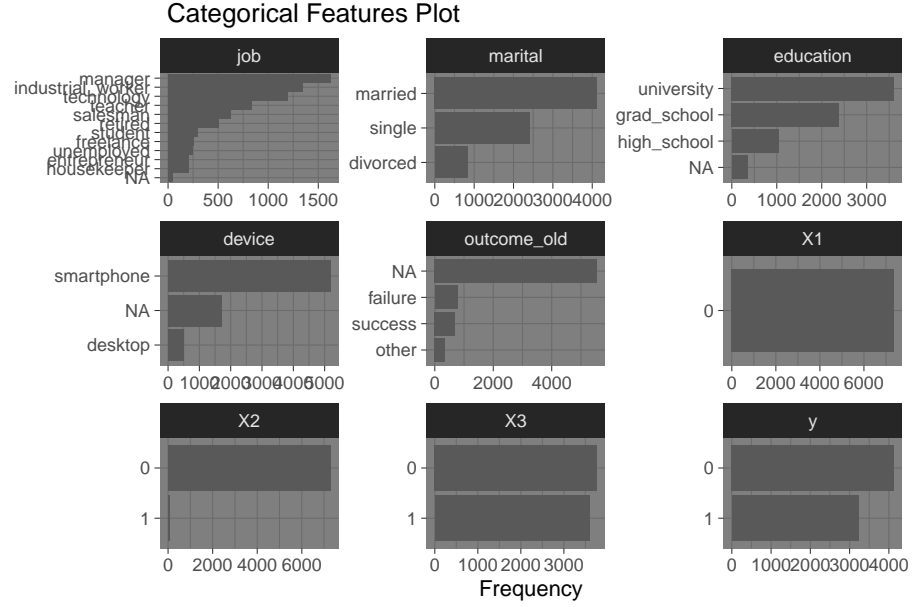


Figure 5: Categorical Features Plot

```
digits = 2,
caption = capt) %>%
kable_styling(latex_options = c("striped", "hold_position"))
```

Table 1: Average and maximum miles per gallon for each number of cylinders class.

cyl	Average	Max	Sqrt
4	26.66	33.9	56.62
6	19.74	21.4	31.08
8	15.10	19.2	54.21

If you want to manually insert the values in the table, you can do it, too (see Table 2).

Table 2: Number of different levels and the number of predictors that have this amount of levels.

	Col 1	Col 2	Col 3	Col 4
Number of different values	2	4	12	> 300
Number of predictors	...	...	...	...

## 3 Models

### 3.1 Linear model

The first approach is to fit a linear model, that is, the regression function  $f$  in (??) is assumed to be of the form

$$f(X) = \beta_0 + \sum_{j=1}^p \beta_j X_j,$$

where  $\beta_j \in \mathbb{R}$  is the coefficient of the  $j$ th predictor.

Fitting this model to the training data, we obtain a predictive model  $\hat{f}_{LM}$ . The training and cross-validation error of this model can be found in Table 3...

There are ... statistically significant predictors...

Possible meaning and interpretation of some predictors

## 4 Validation

We implement a cross-validation ....

## 5 Results

### 5.1 Preliminary Implementation

We started by dividing the set into 75-35 percent split and running them through different machine learning models in a crude manner to check out of the box which model tends to perform best on the given dataset.

```
#Support Vector Machine
rm(list=ls())
# Importing the dataset
dataset = read.csv('train.csv')

dataset$days_elapsed_old[dataset$days_elapsed_old<1] <- 0
dataset[ dataset == "na" ] <- NA

#Factor like columns
dataset$job=as.integer(as.factor(dataset$job))
dataset$marital=as.integer(as.factor(dataset$marital))
dataset$education=as.integer(as.factor(dataset$education))
dataset$device=as.integer(as.factor(dataset$device))
dataset$outcome_old=as.integer(as.factor(dataset$outcome_old))
dataset[is.na(dataset)] <- 0

# Encoding the target feature as factor
dataset$y= factor(dataset$y, levels = c(0, 1))

# Splitting the dataset into the Training set and Test set
#install.packages('caTools')
library(caTools)
set.seed(123)

split = sample.split(dataset$y, SplitRatio = 0.75)

training_set = subset(dataset, split == TRUE)
test_set = subset(dataset, split == FALSE)

# Feature Scaling
training_set[-17] = scale(training_set[-17])
test_set[-17] = scale(test_set[-17])

# Fitting SVM to the Training set
#install.packages('e1071')
```



```

library(e1071)
classifier = svm(formula = y ~ .,
                 data = training_set,
                 type = 'C-classification',
                 kernel = 'radial')

# Predicting the Test set results
y_pred = predict(classifier, newdata = test_set[-17], drop=TRUE)
y_pred

```

```

##      2      5      6      7     11     15     16     18     20     21     31     36     37     41     42     43
##      1      1      1      0      0      0      1      0      1      0      1      1      1      0      0      1
##     56     59     63     64     67     69     83     86     89    101    110    115    124    125    128    140
##      1      0      1      0      1      0      0      0      0      1      0      0      1      0      1      0
##    143    148    149    151    154    155    158    160    163    164    165    167    169    177    179    182
##      0      0      1      0      1      0      1      0      1      1      0      0      1      0      1      0
##    183    187    194    197    201    202    205    211    217    218    220    225    226    229    236    245
##      0      0      0      0      0      0      0      1      1      1      0      1      0      0      1      0
##    247    252    254    256    261    264    267    270    274    279    283    290    291    293    298    301
##      0      0      1      1      0      0      0      1      0      1      0      0      0      0      0      0
##    303    314    321    322    323    324    326    334    339    341    346    349    350    351    353    365
##      1      1      1      1      0      1      0      1      0      0      0      1      0      0      0      0
##    366    367    371    374    383    385    389    397    399    404    405    406    407    413    414    420
##      0      0      0      0      0      1      1      0      1      0      1      1      0      0      1      1
##    421    422    438    439    445    446    449    450    456    464    468    474    476    481    485    499
##      1      0      1      1      0      0      1      0      1      1      1      1      0      1      1      0
##    500    506    509    512    513    520    524    528    531    532    535    537    540    551    552    554
##      0      1      1      0      1      0      0      0      1      0      1      0      0      1      0      1
##    555    559    560    570    573    574    587    589    598    599    603    604    605    607    608    616
##      0      0      0      0      1      1      0      1      1      1      1      0      0      1      1      1
##    622    626    633    634    637    638    649    654    655    656    657    662    666    667    668    669
##      1      1      0      0      1      1      0      0      0      0      0      0      1      0      1      1
##    673    674    675    676    682    683    685    689    692    705    711    717    719    721    725    726
##      0      1      0      1      1      0      0      0      0      0      0      0      1      1      0      1
##    727    729    731    734    737    741    745    751    756    762    766    768    774    775    778    780
##      0      1      0      0      0      0      1      0      1      0      0      0      0      1      1      1
##    784    788    794    796    797    802    803    808    816    818    821    822    827    829    832    838
##      0      0      0      0      1      0      0      0      1      1      0      0      0      0      1      0
##    842    848    849    860    861    865    870    876    878    880    881    887    888    892    893    900
##      1      0      0      1      0      1      1      0      0      0      0      0      1      1      0      0
##    901    907    913    915    919    922    923    925    927    931    936    944    951    953    954    958
##      1      0      0      0      1      0      0      0      0      0      1      1      0      0      0      1
##    959    960    962    964    967    971    982    998    999   1003   1008   1013   1026   1028   1034   1042
##      0      1      0      1      1      0      0      0      0      1      0      1      1      0      1      0
##   1047   1049   1050   1056   1062   1069   1070   1081   1085   1089   1093   1096   1111   1119   1129   1131
##      0      1      1      1      0      0      0      0      0      0      1      1      1      0      1      1
##   1133   1134   1138   1140   1142   1146   1147   1148   1153   1156   1162   1167   1170   1175   1176   1182
##      0      0      1      1      0      0      0      0      0      0      1      1      0      0      0      0
##   1185   1193   1198   1202   1213   1214   1225   1232   1240   1241   1244   1249   1252   1257   1262   1266
##      0      1      0      1      1      1      0      1      0      0      0      0      1      1      0      0
##   1269   1270   1278   1284   1287   1290   1294   1303   1305   1313   1318   1319   1329   1330   1335   1338
##      0      1      0      0      1      1      1      0      1      1      1      0      1      0      0      1
##   1339   1341   1344   1347   1349   1356   1357   1364   1367   1368   1373   1374   1375   1379   1383   1385
##      0      1      1      0      1      0      0      0      1      1      0      0      0      0      0      0

```

##	1389	1396	1404	1406	1407	1408	1410	1412	1419	1421	1423	1427	1428	1429	1432	1436
##	1	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1
##	1443	1446	1451	1453	1457	1463	1466	1473	1475	1476	1477	1478	1479	1481	1483	1484
##	0	0	1	1	0	0	1	0	1	1	0	0	0	1	1	1
##	1485	1487	1492	1496	1499	1501	1507	1508	1512	1516	1520	1523	1527	1529	1533	1546
##	0	0	1	0	0	1	0	0	0	1	1	1	1	0	1	0
##	1547	1548	1552	1553	1558	1561	1565	1569	1571	1577	1580	1581	1599	1619	1621	1623
##	0	1	0	0	0	1	1	1	1	0	1	1	1	0	1	0
##	1634	1638	1640	1641	1645	1652	1665	1673	1677	1683	1689	1694	1718	1719	1723	1727
##	1	1	0	1	0	0	1	1	1	0	0	0	1	0	0	0
##	1729	1730	1731	1735	1736	1737	1740	1743	1746	1751	1752	1754	1760	1761	1763	1767
##	1	0	0	1	0	1	1	0	1	1	0	0	0	1	1	1
##	1769	1782	1783	1792	1796	1803	1806	1807	1808	1816	1821	1823	1824	1827	1830	1837
##	0	0	0	1	0	0	1	1	0	0	0	0	1	0	0	0
##	1839	1840	1843	1846	1849	1855	1856	1858	1864	1866	1870	1872	1876	1878	1882	1891
##	0	0	0	0	1	0	0	0	0	0	0	1	1	0	1	0
##	1893	1900	1904	1925	1926	1928	1929	1933	1939	1943	1956	1964	1966	1972	1975	1976
##	1	1	1	0	1	0	0	1	1	0	1	0	0	1	0	0
##	1981	1991	1999	2003	2004	2007	2008	2015	2017	2018	2020	2027	2028	2029	2035	2042
##	0	1	1	1	0	0	0	0	0	0	0	0	1	0	1	0
##	2046	2049	2050	2052	2053	2056	2073	2074	2076	2081	2083	2084	2088	2093	2098	2102
##	1	0	0	0	0	0	0	0	1	1	1	0	0	0	0	0
##	2107	2113	2115	2117	2120	2134	2135	2139	2140	2145	2150	2152	2154	2155	2162	2167
##	1	1	0	0	0	0	1	0	1	1	0	0	1	1	0	1
##	2168	2174	2175	2188	2198	2199	2203	2216	2217	2218	2222	2226	2230	2235	2241	2247
##	0	1	1	0	0	1	0	1	1	0	0	1	1	0	1	0
##	2253	2254	2255	2256	2259	2264	2270	2272	2275	2277	2278	2282	2283	2289	2309	2315
##	1	1	0	0	1	0	0	1	0	0	0	0	0	1	1	0
##	2316	2322	2327	2329	2330	2332	2338	2345	2349	2353	2356	2357	2365	2373	2376	2377
##	1	0	0	0	0	0	1	1	0	0	1	0	0	0	0	0
##	2385	2389	2391	2392	2394	2400	2403	2406	2408	2412	2421	2426	2430	2431	2432	2436
##	1	0	1	1	0	0	0	0	1	0	1	0	1	1	1	1
##	2439	2447	2457	2460	2462	2464	2465	2466	2469	2474	2475	2479	2486	2488	2491	2504
##	0	0	1	1	1	0	0	0	0	0	1	1	1	0	1	0
##	2505	2510	2514	2515	2516	2520	2521	2523	2526	2531	2534	2535	2537	2538	2541	2542
##	0	1	0	1	0	0	0	0	0	0	1	1	0	0	1	1
##	2546	2554	2563	2574	2579	2580	2581	2582	2586	2591	2594	2595	2598	2601	2610	2615
##	0	0	1	1	1	0	0	0	0	0	1	1	1	0	0	0
##	2619	2626	2627	2631	2632	2639	2640	2644	2645	2647	2649	2656	2657	2658	2659	2661
##	1	0	0	0	1	0	0	0	1	0	1	0	1	0	0	1
##	2662	2669	2672	2675	2679	2688	2693	2695	2696	2699	2703	2704	2712	2722	2736	2744
##	1	0	0	1	0	0	0	1	0	1	1	1	1	0	0	1
##	2746	2747	2751	2753	2754	2755	2758	2763	2767	2768	2769	2770	2771	2780	2784	2785
##	1	1	0	0	0	1	0	0	0	1	0	0	0	1	0	1
##	2787	2793	2798	2803	2804	2807	2810	2812	2820	2825	2832	2834	2839	2843	2848	2850
##	0	0	1	1	1	0	0	0	0	0	0	1	0	1	0	1
##	2856	2866	2872	2873	2878	2879	2883	2892	2899	2900	2902	2905	2906	2915	2916	2923
##	0	0	0	0	1	1	1	1	1	0	0	1	0	1	0	0
##	2927	2934	2936	2940	2943	2945	2949	2951	2957	2959	2962	2963	2965	2967	2981	2984
##	1	1	0	1	1	0	0	1	0	0	1	0	0	1	1	0
##	2985	2989	3004	3011	3017	3028	3037	3042	3044	3050	3058	3059	3061	3064	3067	3068
##	0	1	1	0	1	1	1	0	1	0	0	0	0	0	0	1
##	3071	3074	3075	3076	3082	3089	3093	3095	3098	3101	3102	3105	3106	3111	3119	3127
##	1	0	0	1	1	0	0	0	0	0	1	1	1	0	1	0

##	3134	3136	3137	3147	3151	3160	3172	3173	3180	3181	3185	3186	3189	3193	3196	3198
##	0	0	1	0	1	0	1	0	0	1	0	0	0	0	0	1
##	3204	3206	3210	3212	3217	3227	3231	3233	3238	3240	3243	3245	3246	3249	3254	3257
##	1	1	0	1	0	0	0	0	1	1	0	0	0	1	0	0
##	3259	3265	3272	3282	3285	3289	3294	3305	3310	3323	3325	3328	3329	3330	3339	3341
##	1	0	1	0	1	1	1	0	0	0	0	0	0	1	1	0
##	3344	3348	3350	3354	3356	3357	3359	3363	3364	3374	3384	3387	3395	3399	3402	3405
##	0	0	1	0	0	1	0	0	1	1	0	0	0	0	0	1
##	3408	3409	3416	3417	3419	3427	3430	3431	3432	3434	3435	3438	3441	3450	3461	3462
##	0	0	0	1	0	1	0	0	0	0	0	1	1	0	1	1
##	3464	3466	3467	3477	3478	3483	3490	3493	3499	3503	3506	3510	3514	3523	3524	3533
##	1	1	1	0	0	1	0	0	0	1	0	0	0	0	0	1
##	3534	3541	3542	3544	3545	3548	3549	3551	3558	3559	3563	3579	3580	3583	3585	3587
##	0	1	0	0	1	0	1	1	1	0	0	1	1	1	1	1
##	3590	3591	3592	3596	3597	3598	3602	3603	3608	3610	3613	3620	3621	3624	3626	3630
##	1	0	1	1	0	0	0	1	1	1	1	0	0	0	0	1
##	3631	3641	3646	3647	3650	3654	3659	3660	3664	3676	3689	3692	3695	3700	3706	3707
##	0	0	0	0	0	1	1	1	0	0	1	0	1	1	0	1
##	3711	3714	3715	3724	3726	3727	3741	3742	3749	3755	3756	3758	3761	3763	3766	3775
##	1	0	0	0	1	1	1	0	0	1	0	0	0	0	1	0
##	3778	3779	3781	3782	3789	3795	3801	3808	3819	3823	3824	3833	3836	3839	3841	3846
##	1	1	1	1	0	1	0	0	0	0	1	0	0	1	0	1
##	3847	3851	3872	3877	3878	3881	3882	3883	3893	3895	3896	3898	3899	3900	3906	3914
##	0	0	1	1	1	0	1	0	1	0	0	0	0	0	0	0
##	3917	3924	3927	3928	3937	3938	3941	3949	3951	3952	3958	3960	3962	3963	3965	3967
##	0	1	0	1	1	0	0	0	1	1	0	0	0	1	1	0
##	3969	3970	3972	3973	3974	3981	3982	3993	3996	3997	4002	4004	4006	4007	4015	4017
##	0	1	0	1	0	1	0	0	0	0	1	0	1	0	1	0
##	4018	4021	4033	4034	4047	4048	4056	4060	4065	4066	4069	4075	4076	4077	4078	4083
##	1	0	1	0	0	0	0	0	0	1	0	0	0	0	0	0
##	4088	4090	4092	4095	4097	4100	4108	4109	4110	4112	4113	4122	4127	4131	4134	4135
##	0	1	1	0	0	1	0	0	1	1	1	0	1	1	0	0
##	4136	4138	4139	4140	4150	4151	4152	4157	4158	4162	4166	4167	4169	4171	4175	4178
##	1	0	0	0	0	0	0	0	0	0	0	0	1	1	1	0
##	4184	4185	4190	4191	4197	4204	4205	4208	4209	4210	4212	4215	4217	4218	4219	4225
##	0	0	0	1	0	1	0	1	1	0	0	0	0	1	1	1
##	4236	4238	4241	4243	4245	4247	4248	4249	4256	4259	4266	4269	4271	4280	4282	4283
##	1	1	1	0	1	1	0	0	0	1	1	0	1	0	1	0
##	4287	4290	4291	4293	4317	4326	4329	4333	4334	4337	4340	4345	4353	4359	4364	4366
##	0	1	0	0	0	0	0	1	0	0	0	0	1	0	0	1
##	4371	4374	4377	4378	4379	4381	4388	4390	4391	4396	4403	4407	4409	4413	4415	4417
##	0	1	0	0	1	0	1	0	0	0	0	0	1	0	0	0
##	4419	4423	4424	4427	4428	4431	4433	4438	4439	4440	4444	4445	4448	4456	4461	4464
##	1	1	1	0	0	0	0	0	0	0	1	1	0	0	1	0
##	4469	4479	4481	4482	4484	4485	4486	4489	4492	4498	4502	4505	4507	4509	4510	4512
##	0	0	0	0	1	0	0	0	0	0	1	0	0	0	1	0
##	4517	4520	4523	4524	4526	4533	4534	4540	4542	4548	4552	4560	4572	4578	4582	4585
##	1	0	1	0	0	0	1	1	1	0	0	1	1	1	1	0
##	4596	4625	4630	4633	4642	4654	4656	4670	4674	4678	4688	4693	4698	4703	4715	4716
##	1	0	1	1	1	0	0	0	1	1	1	0	1	1	0	1
##	4722	4723	4731	4739	4742	4746	4748	4753	4757	4759	4763	4771	4775	4778	4780	4783
##	0	1	0	0	1	0	1	1	0	1	0	1	0	0	0	0
##	4784	4786	4787	4788	4801	4806	4807	4811	4817	4826	4831	4840	4845	4854	4858	4863
##	1	1	0	0	1	1	0	1	0	0	0	1	1	0	1	1

##	4868	4870	4871	4877	4879	4889	4890	4891	4899	4902	4905	4906	4908	4913	4917	4919
##	0	0	1	1	0	0	0	0	1	0	0	0	0	0	0	0
##	4924	4925	4926	4927	4935	4936	4937	4938	4939	4940	4946	4947	4950	4959	4965	4966
##	1	0	0	0	0	1	1	1	0	1	0	0	0	0	0	1
##	4968	4970	4972	4978	4980	4984	4987	4988	4989	4991	5000	5003	5006	5009	5011	5014
##	0	0	0	0	0	1	0	0	1	0	0	0	0	0	1	0
##	5016	5017	5020	5021	5029	5039	5041	5051	5056	5059	5064	5070	5074	5075	5087	5089
##	1	0	1	0	1	0	0	0	1	1	0	0	1	1	0	0
##	5093	5095	5100	5101	5116	5123	5124	5127	5131	5133	5134	5137	5138	5140	5144	5153
##	0	0	1	1	1	1	1	0	1	1	0	1	1	1	0	1
##	5154	5163	5166	5172	5186	5188	5189	5193	5194	5197	5204	5206	5211	5214	5215	5216
##	1	0	0	0	1	0	0	1	0	0	0	1	1	0	0	0
##	5218	5226	5227	5230	5236	5244	5245	5248	5250	5251	5254	5261	5262	5264	5265	5268
##	1	0	1	1	1	1	1	0	0	1	0	0	1	1	1	0
##	5269	5273	5275	5276	5280	5283	5284	5294	5295	5296	5300	5303	5305	5310	5330	5335
##	0	0	0	0	1	1	1	0	1	1	0	0	1	1	0	0
##	5352	5353	5355	5357	5358	5359	5360	5363	5365	5366	5369	5371	5376	5378	5382	5385
##	0	0	1	0	1	1	0	0	1	0	0	0	0	1	0	1
##	5389	5393	5400	5402	5405	5406	5407	5410	5411	5417	5418	5419	5431	5432	5433	5442
##	1	1	1	0	1	0	1	0	0	0	0	1	0	1	1	1
##	5449	5452	5454	5455	5462	5466	5472	5473	5474	5483	5491	5495	5503	5508	5510	5511
##	0	0	1	0	0	0	1	1	1	1	0	1	1	0	1	1
##	5513	5515	5520	5524	5529	5539	5545	5550	5551	5554	5566	5584	5589	5599	5602	5604
##	0	0	1	1	1	0	0	0	0	1	0	0	0	0	1	0
##	5607	5608	5614	5616	5621	5622	5628	5638	5642	5651	5654	5662	5666	5675	5676	5691
##	1	1	0	0	1	1	0	1	1	0	0	0	0	0	1	0
##	5693	5704	5707	5709	5710	5715	5718	5720	5723	5726	5729	5732	5736	5739	5740	5745
##	1	0	0	1	0	1	0	1	0	1	0	1	0	0	0	1
##	5746	5751	5752	5757	5758	5759	5765	5770	5772	5777	5779	5780	5782	5797	5806	5810
##	1	0	0	1	0	0	0	1	1	0	1	0	0	1	0	0
##	5811	5814	5820	5828	5835	5849	5857	5862	5865	5867	5868	5871	5875	5878	5881	5882
##	1	0	0	0	0	1	1	0	0	1	0	1	0	0	0	0
##	5887	5893	5901	5904	5909	5911	5922	5923	5924	5925	5930	5931	5932	5934	5937	5941
##	0	1	0	0	0	0	0	0	1	0	0	1	1	0	1	0
##	5946	5949	5951	5953	5963	5973	5977	5978	5980	5984	5993	5995	5998	6000	6004	6008
##	1	0	1	1	0	0	0	1	1	1	0	0	1	1	0	0
##	6015	6016	6018	6030	6035	6044	6047	6049	6050	6052	6056	6057	6059	6063	6065	6066
##	1	1	0	0	0	1	1	0	1	1	0	0	1	0	0	0
##	6068	6069	6076	6096	6097	6099	6100	6103	6106	6109	6115	6116	6117	6119	6121	6125
##	0	0	1	1	0	1	1	0	1	1	0	1	0	0	0	1
##	6129	6142	6143	6146	6148	6158	6159	6167	6168	6169	6177	6185	6190	6193	6196	6199
##	1	0	0	1	0	0	1	1	0	1	1	1	0	0	0	1
##	6200	6206	6211	6219	6221	6222	6223	6224	6228	6235	6237	6242	6246	6250	6258	6264
##	1	1	0	1	1	0	1	0	0	0	1	0	0	0	0	0
##	6265	6268	6272	6275	6284	6285	6290	6293	6294	6296	6306	6317	6324	6325	6328	6334
##	0	0	0	0	0	1	0	0	0	0	1	0	1	1	0	1
##	6337	6340	6344	6348	6350	6352	6353	6354	6358	6360	6361	6362	6363	6379	6380	6391
##	0	1	0	0	1	0	0	1	0	1	0	1	1	0	0	0
##	6393	6397	6400	6411	6415	6418	6419	6422	6423	6430	6437	6440	6444	6451	6460	6462
##	0	1	1	1	0	0	0	1	1	0	0	0	0	1	0	0
##	6463	6465	6466	6467	6469	6471	6477	6478	6481	6483	6485	6486	6487	6488	6489	6490
##	0	1	0	1	1	0	1	0	1	0	0	0	1	0	0	0
##	6491	6492	6497	6499	6502	6508	6515	6521	6522	6527	6532	6534	6535	6538	6549	6551
##	0	0	0	0	1	0	1	1	0	1	1	1	1	1	1	0

##	6553	6555	6559	6561	6563	6565	6566	6569	6575	6580	6582	6586	6614	6615	6617	6618
##	1	0	0	1	0	1	0	1	0	1	1	0	0	1	0	0
##	6620	6632	6646	6654	6658	6661	6663	6664	6665	6673	6677	6681	6682	6686	6687	6690
##	0	0	0	0	0	0	0	1	1	0	0	1	1	0	0	1
##	6693	6694	6695	6697	6698	6702	6705	6717	6724	6727	6728	6730	6732	6733	6738	6740
##	1	0	1	1	0	0	1	1	1	0	0	1	1	1	1	1
##	6743	6751	6756	6757	6760	6763	6767	6770	6771	6777	6782	6783	6784	6788	6790	6793
##	1	0	0	0	1	1	0	0	0	1	0	1	0	0	0	1
##	6797	6798	6803	6807	6813	6815	6825	6832	6837	6848	6850	6852	6855	6859	6864	6868
##	0	0	1	0	1	0	0	1	1	0	0	0	0	0	1	0
##	6871	6873	6885	6887	6891	6895	6900	6903	6909	6920	6922	6926	6930	6932	6941	6952
##	1	1	1	0	0	1	0	0	1	0	0	0	0	0	0	0
##	6953	6958	6961	6964	6974	6976	6977	6978	6979	6981	6986	6993	7002	7005	7015	7020
##	0	1	0	0	0	0	0	0	1	1	0	1	0	1	0	1
##	7022	7026	7028	7039	7043	7047	7058	7073	7074	7081	7083	7084	7085	7086	7087	7089
##	0	1	0	0	0	0	1	1	1	0	0	0	0	1	0	1
##	7094	7097	7098	7104	7110	7115	7118	7124	7129	7133	7138	7146	7155	7160	7164	7165
##	1	0	0	0	1	1	1	0	1	1	1	0	0	0	1	0
##	7169	7172	7173	7180	7187	7194	7196	7200	7202	7203	7204	7211	7215	7217	7220	7226
##	1	0	0	1	0	0	0	0	1	0	0	0	1	0	0	0
##	7237	7240	7241	7244	7248	7249	7254	7263	7264	7265	7267	7268	7271	7273	7274	7277
##	1	1	0	0	0	1	1	1	0	0	0	0	1	0	0	0
##	7278	7283	7289	7295	7298	7300	7311	7322	7325	7326	7327	7332	7337	7343	7344	7345
##	1	0	0	0	1	0	0	1	1	0	0	1	1	1	0	0
##	7347	7348	7351	7355	7357	7363	7366	7367	7368	7372	7374	7380	7383	7389	7391	7393
##	0	0	0	1	1	0	1	0	0	0	1	0	0	0	0	0
##	7396	7402	7405	7409	7415	7418	7425	7428	7433	7435	7438	7439	7445	7448	7457	7458
##	1	0	1	0	1	0	0	0	1	0	1	0	1	1	0	0
##	7463	7465	7470	7471	7475	7476	7483	7484	7487	7489	7491	7493	7494	7495	7498	7504
##	0	1	0	1	0	0	1	0	1	0	0	1	0	0	0	1
##	7508	7509	7521	7523	7524	7527	7538	7542	7546	7547	7553	7560	7570	7571	7578	7579
##	0	1	1	1	0	1	0	1	1	1	0	0	0	0	0	1
##	7583	7584	7593	7597	7603	7606	7608	7609	7615	7618	7620	7628	7630	7631	7637	7638
##	0	0	1	1	1	0	1	0	0	0	0	1	0	0	0	0
##	7640	7642	7644	7649	7653	7663	7668	7669	7671	7672	7673	7677	7682	7684	7686	7687
##	0	1	1	1	0	0	0	1	0	0	1	0	1	0	0	1
##	7691	7694	7698	7701	7702	7718	7721	7722	7724	7725	7738	7742	7744	7745	7747	7756
##	1	0	0	0	0	0	0	0	1	0	1	1	1	1	1	1
##	7760	7767	7768	7771	7772	7774	7783	7785	7787	7788	7793	7798	7799	7801	7805	7808
##	0	1	0	1	1	1	0	1	1	1	0	0	0	1	1	0
##	7813	7815	7821	7824	7829	7832	7839	7840	7845	7847	7850	7851	7865	7868	7874	7879
##	1	0	0	0	0	0	1	1	1	0	1	0	0	0	1	0
##	7881	7882	7884	7886	7889	7890	7893	7902	7904	7908	7912	7918	7920	7921	7925	7930
##	1	0	0	0	1	0	0	1	1	0	1	0	0	0	1	0
##	7931	7937	7938	7947	7948	7952	7960	7962	7970	7972	7973	7975	7978	7979	7980	7983
##	0	0	0	1	1	0	1	1	0	0	1	0	0	1	1	1
##	7985	7987	7988	7991	8000	8002	8015	8019	8021	8025	8033	8036	8045	8051	8052	8056
##	0	1	1	1	0	1	1	0	0	1	1	0	1	1	0	0
##	8057	8059	8063	8067	8078	8079	8080	8083	8087	8091	8096	8099	8104	8108	8109	8112
##	1	0	1	0	1	0	0	0	0	0	0	0	0	0	1	0
##	8124	8136	8138	8140	8142	8144	8147	8149	8153	8154	8157	8160	8170	8176	8183	8206
##	0	0	0	1	0	0	0	0	1	1	1	1	0	1	1	0
##	8207	8209	8211	8216	8230	8241	8245	8249	8255	8265	8266	8267	8273	8284	8285	8289
##	0	1	0	0	0	0	0	0	0	0	0	1	1	1	1	0

```
## 8294 8297 8303 8305 8319 8320 8322 8325 8330 8335 8339 8340 8341 8349 8357 8358
##      0      0      0      1      1      1      1      0      1      1      0      1      1      0      1      0
## 8366 8371 8372 8373 8374 8386 8389 8396 8399 8412 8413 8425 8429 8434 8438 8440
##      0      0      1      0      1      0      0      0      0      0      0      0      1      1      1      1
## 8446 8451 8453 8454 8462 8464 8469 8472 8473 8479 8481 8487 8491 8494 8499 8501
##      0      1      0      1      0      1      0      0      0      1      1      0      0      1      1      0
## 8504 8514 8521 8526
##      1      0      1      0
## Levels: 0 1
```

```
# Making the Confusion Matrix
```

```
cm = table(test_set[, 17], y_pred)
cm
```

```
##      y_pred
##           0      1
## 0 1083   157
## 1   168   724
```

```
print("=====SVM=====")
```

```
## [1] "=====SVM====="
```

```
library(ggplot2)
library(lattice)
library(caret)
confusionMatrix(cm)
```

```
## Confusion Matrix and Statistics
```

```
##
##      y_pred
##           0      1
## 0 1083   157
## 1   168   724
##
##              Accuracy : 0.8476
##              95% CI : (0.8316, 0.8626)
##      No Information Rate : 0.5868
##      P-Value [Acc > NIR] : <2e-16
##
##              Kappa : 0.6862
##
##  Mcnemar's Test P-Value : 0.5791
##
##              Sensitivity : 0.8657
##              Specificity : 0.8218
##      Pos Pred Value : 0.8734
##      Neg Pred Value : 0.8117
##              Prevalence : 0.5868
##      Detection Rate : 0.5080
##      Detection Prevalence : 0.5816
##      Balanced Accuracy : 0.8438
##
##      'Positive' Class : 0
##
```

```

#####
rm(list=ls())
#Random Forest Classification

# Importing the dataset
dataset = read.csv('train.csv')

dataset$days_elapsed_old[dataset$days_elapsed_old<1] <- 0
dataset[ dataset == "na" ] <- NA

#Factor like columns
dataset$job=as.integer(as.factor(dataset$job))
dataset$marital=as.integer(as.factor(dataset$marital))
dataset$education=as.integer(as.factor(dataset$education))
dataset$device=as.integer(as.factor(dataset$device))
dataset$outcome_old=as.integer(as.factor(dataset$outcome_old))
dataset[is.na(dataset)] <- 0

# Encoding the target feature as factor
dataset$y = factor(dataset$y, levels = c(0, 1))

# Splitting the dataset into the Training set and Test set
# install.packages('caTools')
library(caTools)
set.seed(123)
split = sample.split(dataset$y, SplitRatio = 0.75)
training_set = subset(dataset, split == TRUE)
test_set = subset(dataset, split == FALSE)

# Feature Scaling #for higher resolution visualisation only we are using feature scaling,RF doesnt need
training_set[-17] = scale(training_set[-17])
test_set[-17] = scale(test_set[-17])

# Fitting Random Forest Classification to the Training set
#install.packages('randomForest')
library(randomForest)
classifier = randomForest(x = training_set[-17],
                          y = training_set$y)#, ntree = 700)

# Predicting the Test set results
y_pred = predict(classifier, newdata = test_set[-17])
y_pred

```

##	2	5	6	7	11	15	16	18	20	21	31	36	37	41	42	43
##	1	1	1	0	0	0	0	0	1	0	0	1	1	0	0	1
##	56	59	63	64	67	69	83	86	89	101	110	115	124	125	128	140
##	1	0	1	0	1	0	0	0	0	1	0	0	1	1	0	0
##	143	148	149	151	154	155	158	160	163	164	165	167	169	177	179	182
##	0	0	1	0	1	0	1	0	1	1	0	1	1	0	0	0
##	183	187	194	197	201	202	205	211	217	218	220	225	226	229	236	245
##	0	0	0	0	0	1	0	1	1	1	0	1	0	0	1	0
##	247	252	254	256	261	264	267	270	274	279	283	290	291	293	298	301
##	0	0	1	1	0	0	0	1	0	1	0	0	0	0	0	0
##	303	314	321	322	323	324	326	334	339	341	346	349	350	351	353	365

##	1	1	1	1	1	1	0	1	1	0	0	1	0	0	0	0
##	366	367	371	374	383	385	389	397	399	404	405	406	407	413	414	420
##	0	0	0	0	1	0	1	0	1	0	1	1	0	0	1	0
##	421	422	438	439	445	446	449	450	456	464	468	474	476	481	485	499
##	1	0	1	1	0	0	1	0	1	1	1	1	0	1	1	0
##	500	506	509	512	513	520	524	528	531	532	535	537	540	551	552	554
##	0	1	1	0	0	0	0	0	1	0	1	0	0	1	0	1
##	555	559	560	570	573	574	587	589	598	599	603	604	605	607	608	616
##	0	0	0	0	1	1	0	1	1	1	1	0	0	1	1	1
##	622	626	633	634	637	638	649	654	655	656	657	662	666	667	668	669
##	1	1	1	0	1	1	0	0	0	0	0	0	1	1	1	1
##	673	674	675	676	682	683	685	689	692	705	711	717	719	721	725	726
##	0	1	0	1	1	0	1	0	0	0	0	0	1	1	0	1
##	727	729	731	734	737	741	745	751	756	762	766	768	774	775	778	780
##	0	1	0	1	0	0	1	0	1	1	0	0	0	0	0	1
##	784	788	794	796	797	802	803	808	816	818	821	822	827	829	832	838
##	0	0	0	0	1	0	0	0	1	0	0	0	0	0	1	0
##	842	848	849	860	861	865	870	876	878	880	881	887	888	892	893	900
##	1	0	0	1	0	1	1	0	0	0	0	0	0	1	0	0
##	901	907	913	915	919	922	923	925	927	931	936	944	951	953	954	958
##	1	0	0	0	1	0	0	0	0	0	1	0	0	0	0	1
##	959	960	962	964	967	971	982	998	999	1003	1008	1013	1026	1028	1034	1042
##	0	1	0	0	1	0	0	0	0	1	0	1	0	0	1	0
##	1047	1049	1050	1056	1062	1069	1070	1081	1085	1089	1093	1096	1111	1119	1129	1131
##	0	0	0	1	0	0	0	0	0	0	1	1	0	0	1	1
##	1133	1134	1138	1140	1142	1146	1147	1148	1153	1156	1162	1167	1170	1175	1176	1182
##	1	0	1	1	0	0	0	0	0	0	1	1	0	0	0	0
##	1185	1193	1198	1202	1213	1214	1225	1232	1240	1241	1244	1249	1252	1257	1262	1266
##	0	1	0	1	1	1	0	1	0	0	0	1	1	1	0	0
##	1269	1270	1278	1284	1287	1290	1294	1303	1305	1313	1318	1319	1329	1330	1335	1338
##	0	1	0	0	1	1	1	0	1	1	1	0	1	0	0	1
##	1339	1341	1344	1347	1349	1356	1357	1364	1367	1368	1373	1374	1375	1379	1383	1385
##	0	1	1	0	1	0	0	0	1	1	0	0	0	1	1	0
##	1389	1396	1404	1406	1407	1408	1410	1412	1419	1421	1423	1427	1428	1429	1432	1436
##	1	1	0	0	0	1	0	1	0	0	0	1	0	1	0	1
##	1443	1446	1451	1453	1457	1463	1466	1473	1475	1476	1477	1478	1479	1481	1483	1484
##	0	0	1	1	0	0	1	0	1	1	0	1	0	1	1	0
##	1485	1487	1492	1496	1499	1501	1507	1508	1512	1516	1520	1523	1527	1529	1533	1546
##	0	0	1	0	1	1	1	0	0	1	1	1	1	0	1	0
##	1547	1548	1552	1553	1558	1561	1565	1569	1571	1577	1580	1581	1599	1619	1621	1623
##	0	1	0	0	0	1	1	1	0	0	1	1	0	0	1	0
##	1634	1638	1640	1641	1645	1652	1665	1673	1677	1683	1689	1694	1718	1719	1723	1727
##	1	1	0	1	0	0	1	1	1	0	0	0	1	0	0	0
##	1729	1730	1731	1735	1736	1737	1740	1743	1746	1751	1752	1754	1760	1761	1763	1767
##	1	0	0	1	0	1	1	0	1	1	0	0	0	1	0	1
##	1769	1782	1783	1792	1796	1803	1806	1807	1808	1816	1821	1823	1824	1827	1830	1837
##	0	0	0	1	0	0	1	1	1	1	0	0	0	0	0	0
##	1839	1840	1843	1846	1849	1855	1856	1858	1864	1866	1870	1872	1876	1878	1882	1891
##	0	0	0	0	1	0	0	0	0	0	0	1	1	0	1	0
##	1893	1900	1904	1925	1926	1928	1929	1933	1939	1943	1956	1964	1966	1972	1975	1976
##	1	1	1	0	1	1	1	1	1	1	1	0	0	1	0	0
##	1981	1991	1999	2003	2004	2007	2008	2015	2017	2018	2020	2027	2028	2029	2035	2042
##	0	1	0	1	0	0	0	0	0	0	0	0	1	0	1	0
##	2046	2049	2050	2052	2053	2056	2073	2074	2076	2081	2083	2084	2088	2093	2098	2102



##	1	0	0	0	0	0	0	0	1	1	1	0	0	0	0	0
##	2107	2113	2115	2117	2120	2134	2135	2139	2140	2145	2150	2152	2154	2155	2162	2167
##	1	0	0	0	0	0	1	0	1	1	0	0	1	1	0	1
##	2168	2174	2175	2188	2198	2199	2203	2216	2217	2218	2222	2226	2230	2235	2241	2247
##	0	1	1	0	0	1	0	1	0	0	0	1	1	0	1	0
##	2253	2254	2255	2256	2259	2264	2270	2272	2275	2277	2278	2282	2283	2289	2309	2315
##	1	0	0	1	1	0	0	1	0	0	0	0	0	0	1	0
##	2316	2322	2327	2329	2330	2332	2338	2345	2349	2353	2356	2357	2365	2373	2376	2377
##	1	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0
##	2385	2389	2391	2392	2394	2400	2403	2406	2408	2412	2421	2426	2430	2431	2432	2436
##	1	0	1	1	0	0	0	0	1	0	1	0	1	1	1	1
##	2439	2447	2457	2460	2462	2464	2465	2466	2469	2474	2475	2479	2486	2488	2491	2504
##	0	0	1	1	1	0	0	0	0	0	0	1	1	0	1	0
##	2505	2510	2514	2515	2516	2520	2521	2523	2526	2531	2534	2535	2537	2538	2541	2542
##	0	0	0	1	0	0	0	0	0	0	1	1	0	0	0	1
##	2546	2554	2563	2574	2579	2580	2581	2582	2586	2591	2594	2595	2598	2601	2610	2615
##	0	0	0	1	1	0	0	0	0	1	1	1	1	0	0	0
##	2619	2626	2627	2631	2632	2639	2640	2644	2645	2647	2649	2656	2657	2658	2659	2661
##	1	0	1	0	1	0	0	0	1	0	1	1	1	0	0	1
##	2662	2669	2672	2675	2679	2688	2693	2695	2696	2699	2703	2704	2712	2722	2736	2744
##	1	0	0	1	0	0	0	1	0	1	1	1	1	0	0	1
##	2746	2747	2751	2753	2754	2755	2758	2763	2767	2768	2769	2770	2771	2780	2784	2785
##	1	1	0	0	0	1	0	0	0	1	0	0	0	0	0	1
##	2787	2793	2798	2803	2804	2807	2810	2812	2820	2825	2832	2834	2839	2843	2848	2850
##	0	0	1	1	1	0	1	0	0	0	0	1	0	1	0	1
##	2856	2866	2872	2873	2878	2879	2883	2892	2899	2900	2902	2905	2906	2915	2916	2923
##	0	0	0	0	1	1	1	0	1	0	0	1	0	1	1	1
##	2927	2934	2936	2940	2943	2945	2949	2951	2957	2959	2962	2963	2965	2967	2981	2984
##	1	1	0	1	1	1	0	1	0	0	1	0	0	1	1	0
##	2985	2989	3004	3011	3017	3028	3037	3042	3044	3050	3058	3059	3061	3064	3067	3068
##	0	1	1	0	1	1	1	0	1	0	0	0	0	0	0	1
##	3071	3074	3075	3076	3082	3089	3093	3095	3098	3101	3102	3105	3106	3111	3119	3127
##	1	0	0	0	1	0	0	0	0	0	1	0	1	0	1	0
##	3134	3136	3137	3147	3151	3160	3172	3173	3180	3181	3185	3186	3189	3193	3196	3198
##	0	0	1	0	1	0	1	0	0	1	0	0	0	0	0	1
##	3204	3206	3210	3212	3217	3227	3231	3233	3238	3240	3243	3245	3246	3249	3254	3257
##	1	0	0	1	1	0	0	0	1	1	0	0	0	1	0	0
##	3259	3265	3272	3282	3285	3289	3294	3305	3310	3323	3325	3328	3329	3330	3339	3341
##	1	0	1	0	1	0	1	0	0	0	0	0	0	1	1	0
##	3344	3348	3350	3354	3356	3357	3359	3363	3364	3374	3384	3387	3395	3399	3402	3405
##	0	0	1	0	0	1	0	0	1	1	0	0	0	0	0	1
##	3408	3409	3416	3417	3419	3427	3430	3431	3432	3434	3435	3438	3441	3450	3461	3462
##	0	0	0	1	0	1	0	0	0	0	0	1	1	0	1	1
##	3464	3466	3467	3477	3478	3483	3490	3493	3499	3503	3506	3510	3514	3523	3524	3533
##	1	1	1	1	0	1	0	0	0	1	1	0	0	0	0	1
##	3534	3541	3542	3544	3545	3548	3549	3551	3558	3559	3563	3579	3580	3583	3585	3587
##	0	1	0	0	1	0	1	1	1	0	0	1	1	1	0	1
##	3590	3591	3592	3596	3597	3598	3602	3603	3608	3610	3613	3620	3621	3624	3626	3630
##	1	0	1	1	0	0	1	1	1	1	1	0	0	0	0	1
##	3631	3641	3646	3647	3650	3654	3659	3660	3664	3676	3689	3692	3695	3700	3706	3707
##	1	0	0	0	0	1	1	1	0	0	1	0	1	1	0	1
##	3711	3714	3715	3724	3726	3727	3741	3742	3749	3755	3756	3758	3761	3763	3766	3775
##	1	0	0	0	1	1	1	0	0	1	0	0	0	1	1	0
##	3778	3779	3781	3782	3789	3795	3801	3808	3819	3823	3824	3833	3836	3839	3841	3846

##	1	1	1	1	0	1	0	0	0	0	0	0	0	1	0	1
##	3847	3851	3872	3877	3878	3881	3882	3883	3893	3895	3896	3898	3899	3900	3906	3914
##	0	0	1	1	1	0	1	0	1	0	1	0	0	0	0	0
##	3917	3924	3927	3928	3937	3938	3941	3949	3951	3952	3958	3960	3962	3963	3965	3967
##	0	1	0	1	1	0	0	0	1	1	1	0	0	1	1	0
##	3969	3970	3972	3973	3974	3981	3982	3993	3996	3997	4002	4004	4006	4007	4015	4017
##	0	1	1	1	0	0	0	0	0	0	1	0	1	0	1	0
##	4018	4021	4033	4034	4047	4048	4056	4060	4065	4066	4069	4075	4076	4077	4078	4083
##	1	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
##	4088	4090	4092	4095	4097	4100	4108	4109	4110	4112	4113	4122	4127	4131	4134	4135
##	0	1	1	0	0	0	1	0	1	1	1	0	1	1	0	0
##	4136	4138	4139	4140	4150	4151	4152	4157	4158	4162	4166	4167	4169	4171	4175	4178
##	1	0	0	0	0	1	0	1	0	0	0	0	1	1	1	1
##	4184	4185	4190	4191	4197	4204	4205	4208	4209	4210	4212	4215	4217	4218	4219	4225
##	0	0	0	1	0	0	0	1	1	0	0	0	0	1	1	1
##	4236	4238	4241	4243	4245	4247	4248	4249	4256	4259	4266	4269	4271	4280	4282	4283
##	1	1	1	0	1	1	0	0	0	1	1	0	1	1	1	0
##	4287	4290	4291	4293	4317	4326	4329	4333	4334	4337	4340	4345	4353	4359	4364	4366
##	0	1	0	1	0	0	0	1	0	0	0	1	1	0	0	1
##	4371	4374	4377	4378	4379	4381	4388	4390	4391	4396	4403	4407	4409	4413	4415	4417
##	0	1	0	0	1	0	1	0	1	0	0	0	1	0	0	0
##	4419	4423	4424	4427	4428	4431	4433	4438	4439	4440	4444	4445	4448	4456	4461	4464
##	1	1	1	0	1	0	0	0	0	0	0	1	0	0	1	0
##	4469	4479	4481	4482	4484	4485	4486	4489	4492	4498	4502	4505	4507	4509	4510	4512
##	0	0	0	1	1	0	0	0	0	0	1	0	0	0	1	0
##	4517	4520	4523	4524	4526	4533	4534	4540	4542	4548	4552	4560	4572	4578	4582	4585
##	1	0	1	0	0	0	1	0	1	1	0	1	1	0	1	1
##	4596	4625	4630	4633	4642	4654	4656	4670	4674	4678	4688	4693	4698	4703	4715	4716
##	1	0	1	1	1	1	1	0	1	1	1	0	1	1	0	1
##	4722	4723	4731	4739	4742	4746	4748	4753	4757	4759	4763	4771	4775	4778	4780	4783
##	0	0	0	0	1	0	1	1	0	1	0	1	0	0	0	0
##	4784	4786	4787	4788	4801	4806	4807	4811	4817	4826	4831	4840	4845	4854	4858	4863
##	1	0	0	0	1	1	0	1	0	0	0	1	1	0	1	1
##	4868	4870	4871	4877	4879	4889	4890	4891	4899	4902	4905	4906	4908	4913	4917	4919
##	0	0	0	1	0	0	0	0	1	0	1	0	1	0	0	0
##	4924	4925	4926	4927	4935	4936	4937	4938	4939	4940	4946	4947	4950	4959	4965	4966
##	1	0	1	0	0	0	1	1	0	1	0	0	0	0	0	1
##	4968	4970	4972	4978	4980	4984	4987	4988	4989	4991	5000	5003	5006	5009	5011	5014
##	0	1	0	0	0	1	0	0	1	0	0	0	0	0	1	0
##	5016	5017	5020	5021	5029	5039	5041	5051	5056	5059	5064	5070	5074	5075	5087	5089
##	1	0	1	0	1	0	0	0	1	1	0	1	1	1	0	0
##	5093	5095	5100	5101	5116	5123	5124	5127	5131	5133	5134	5137	5138	5140	5144	5153
##	0	0	1	1	1	1	1	0	1	1	0	1	1	1	0	1
##	5154	5163	5166	5172	5186	5188	5189	5193	5194	5197	5204	5206	5211	5214	5215	5216
##	1	0	0	0	1	0	0	0	0	0	0	1	1	0	0	0
##	5218	5226	5227	5230	5236	5244	5245	5248	5250	5251	5254	5261	5262	5264	5265	5268
##	1	0	1	1	0	1	0	0	0	1	0	0	1	1	1	0
##	5269	5273	5275	5276	5280	5283	5284	5294	5295	5296	5300	5303	5305	5310	5330	5335
##	0	0	0	0	0	1	1	1	1	1	0	0	1	1	0	0
##	5352	5353	5355	5357	5358	5359	5360	5363	5365	5366	5369	5371	5376	5378	5382	5385
##	0	0	1	0	1	1	0	0	1	0	0	0	0	1	0	1
##	5389	5393	5400	5402	5405	5406	5407	5410	5411	5417	5418	5419	5431	5432	5433	5442
##	1	1	1	0	1	0	1	0	0	0	1	1	0	1	1	1
##	5449	5452	5454	5455	5462	5466	5472	5473	5474	5483	5491	5495	5503	5508	5510	5511

##	0	1	1	0	1	0	1	1	1	1	0	1	0	0	1	1
##	5513	5515	5520	5524	5529	5539	5545	5550	5551	5554	5566	5584	5589	5599	5602	5604
##	0	0	1	1	1	0	0	0	0	1	0	0	0	0	1	0
##	5607	5608	5614	5616	5621	5622	5628	5638	5642	5651	5654	5662	5666	5675	5676	5691
##	1	1	0	0	1	1	0	0	1	0	0	0	0	0	1	0
##	5693	5704	5707	5709	5710	5715	5718	5720	5723	5726	5729	5732	5736	5739	5740	5745
##	1	0	0	1	0	0	0	1	0	1	0	1	0	0	0	1
##	5746	5751	5752	5757	5758	5759	5765	5770	5772	5777	5779	5780	5782	5797	5806	5810
##	1	0	1	1	1	0	0	0	1	0	1	0	0	1	0	0
##	5811	5814	5820	5828	5835	5849	5857	5862	5865	5867	5868	5871	5875	5878	5881	5882
##	1	0	0	0	0	1	1	0	0	1	1	1	0	0	0	0
##	5887	5893	5901	5904	5909	5911	5922	5923	5924	5925	5930	5931	5932	5934	5937	5941
##	0	1	0	0	0	0	0	0	1	0	0	1	1	0	1	0
##	5946	5949	5951	5953	5963	5973	5977	5978	5980	5984	5993	5995	5998	6000	6004	6008
##	1	0	1	0	0	0	0	1	1	1	0	0	1	0	0	0
##	6015	6016	6018	6030	6035	6044	6047	6049	6050	6052	6056	6057	6059	6063	6065	6066
##	1	1	0	0	0	1	1	0	1	1	0	0	1	0	0	0
##	6068	6069	6076	6096	6097	6099	6100	6103	6106	6109	6115	6116	6117	6119	6121	6125
##	0	0	1	1	0	1	1	1	1	1	0	0	0	0	0	1
##	6129	6142	6143	6146	6148	6158	6159	6167	6168	6169	6177	6185	6190	6193	6196	6199
##	1	0	0	1	0	0	1	1	0	1	1	1	0	0	0	1
##	6200	6206	6211	6219	6221	6222	6223	6224	6228	6235	6237	6242	6246	6250	6258	6264
##	1	1	0	1	1	0	1	1	0	0	1	0	0	0	0	0
##	6265	6268	6272	6275	6284	6285	6290	6293	6294	6296	6306	6317	6324	6325	6328	6334
##	0	0	1	0	0	1	0	0	1	0	1	0	1	1	0	1
##	6337	6340	6344	6348	6350	6352	6353	6354	6358	6360	6361	6362	6363	6379	6380	6391
##	1	1	0	0	1	0	0	1	0	1	0	1	1	0	0	0
##	6393	6397	6400	6411	6415	6418	6419	6422	6423	6430	6437	6440	6444	6451	6460	6462
##	0	1	1	1	0	0	0	1	1	0	0	0	0	1	0	0
##	6463	6465	6466	6467	6469	6471	6477	6478	6481	6483	6485	6486	6487	6488	6489	6490
##	0	1	0	1	1	0	1	0	1	0	0	0	1	0	0	0
##	6491	6492	6497	6499	6502	6508	6515	6521	6522	6527	6532	6534	6535	6538	6549	6551
##	0	0	0	0	1	0	0	1	0	1	1	1	1	1	1	0
##	6553	6555	6559	6561	6563	6565	6566	6569	6575	6580	6582	6586	6614	6615	6617	6618
##	1	1	0	1	0	1	0	1	0	1	1	0	1	1	0	1
##	6620	6632	6646	6654	6658	6661	6663	6664	6665	6673	6677	6681	6682	6686	6687	6690
##	0	0	0	0	0	0	0	1	1	0	0	1	1	0	0	1
##	6693	6694	6695	6697	6698	6702	6705	6717	6724	6727	6728	6730	6732	6733	6738	6740
##	1	1	1	1	0	0	1	1	1	0	1	1	1	1	1	1
##	6743	6751	6756	6757	6760	6763	6767	6770	6771	6777	6782	6783	6784	6788	6790	6793
##	1	0	0	0	1	1	0	0	0	1	1	1	0	0	0	0
##	6797	6798	6803	6807	6813	6815	6825	6832	6837	6848	6850	6852	6855	6859	6864	6868
##	0	0	1	0	1	0	0	1	1	0	0	0	0	0	1	0
##	6871	6873	6885	6887	6891	6895	6900	6903	6909	6920	6922	6926	6930	6932	6941	6952
##	1	1	1	0	0	1	0	0	1	0	0	0	0	1	0	0
##	6953	6958	6961	6964	6974	6976	6977	6978	6979	6981	6986	6993	7002	7005	7015	7020
##	0	1	0	0	0	0	0	0	1	1	0	1	0	1	1	1
##	7022	7026	7028	7039	7043	7047	7058	7073	7074	7081	7083	7084	7085	7086	7087	7089
##	0	0	0	0	0	0	1	0	1	0	0	0	0	1	0	1
##	7094	7097	7098	7104	7110	7115	7118	7124	7129	7133	7138	7146	7155	7160	7164	7165
##	1	0	0	0	1	1	1	0	1	1	1	0	0	0	1	0
##	7169	7172	7173	7180	7187	7194	7196	7200	7202	7203	7204	7211	7215	7217	7220	7226
##	1	0	0	0	0	0	1	0	1	0	0	0	1	0	0	0
##	7237	7240	7241	7244	7248	7249	7254	7263	7264	7265	7267	7268	7271	7273	7274	7277

```

##      1      1      1      0      1      1      1      1      1      0      1      0      1      0      0      0
## 7278 7283 7289 7295 7298 7300 7311 7322 7325 7326 7327 7332 7337 7343 7344 7345
##      1      0      0      0      1      0      0      1      1      0      0      1      1      1      0      0
## 7347 7348 7351 7355 7357 7363 7366 7367 7368 7372 7374 7380 7383 7389 7391 7393
##      0      0      0      1      1      0      1      0      0      0      1      0      1      0      0      0
## 7396 7402 7405 7409 7415 7418 7425 7428 7433 7435 7438 7439 7445 7448 7457 7458
##      1      0      1      0      1      0      0      0      1      0      0      0      1      1      0      0
## 7463 7465 7470 7471 7475 7476 7483 7484 7487 7489 7491 7493 7494 7495 7498 7504
##      0      1      0      1      0      0      1      0      1      0      0      1      0      0      0      1
## 7508 7509 7521 7523 7524 7527 7538 7542 7546 7547 7553 7560 7570 7571 7578 7579
##      0      1      0      1      0      1      0      1      1      1      0      0      0      0      0      1
## 7583 7584 7593 7597 7603 7606 7608 7609 7615 7618 7620 7628 7630 7631 7637 7638
##      0      0      0      1      1      0      1      0      0      0      0      1      0      0      0      0
## 7640 7642 7644 7649 7653 7663 7668 7669 7671 7672 7673 7677 7682 7684 7686 7687
##      0      1      1      1      0      0      0      1      0      0      1      0      1      0      0      1
## 7691 7694 7698 7701 7702 7718 7721 7722 7724 7725 7738 7742 7744 7745 7747 7756
##      1      0      0      0      0      0      0      0      1      0      1      1      1      1      1      1
## 7760 7767 7768 7771 7772 7774 7783 7785 7787 7788 7793 7798 7799 7801 7805 7808
##      0      1      1      1      1      1      0      0      1      1      0      0      0      1      1      1
## 7813 7815 7821 7824 7829 7832 7839 7840 7845 7847 7850 7851 7865 7868 7874 7879
##      0      1      0      0      0      1      1      1      1      0      1      0      1      0      1      0
## 7881 7882 7884 7886 7889 7890 7893 7902 7904 7908 7912 7918 7920 7921 7925 7930
##      1      0      1      0      1      0      0      1      1      0      1      0      0      0      1      0
## 7931 7937 7938 7947 7948 7952 7960 7962 7970 7972 7973 7975 7978 7979 7980 7983
##      0      0      0      1      1      0      1      1      0      0      1      0      0      1      1      1
## 7985 7987 7988 7991 8000 8002 8015 8019 8021 8025 8033 8036 8045 8051 8052 8056
##      0      1      1      1      0      1      1      0      0      1      1      0      1      1      0      0
## 8057 8059 8063 8067 8078 8079 8080 8083 8087 8091 8096 8099 8104 8108 8109 8112
##      1      0      1      0      1      0      0      0      0      0      0      0      0      0      1      0
## 8124 8136 8138 8140 8142 8144 8147 8149 8153 8154 8157 8160 8170 8176 8183 8206
##      0      0      0      1      0      0      0      0      1      1      1      1      0      1      1      0
## 8207 8209 8211 8216 8230 8241 8245 8249 8255 8265 8266 8267 8273 8284 8285 8289
##      0      1      0      0      0      0      0      0      0      0      0      1      1      1      1      0
## 8294 8297 8303 8305 8319 8320 8322 8325 8330 8335 8339 8340 8341 8349 8357 8358
##      0      0      0      1      1      1      1      0      1      1      0      1      1      0      0      0
## 8366 8371 8372 8373 8374 8386 8389 8396 8399 8412 8413 8425 8429 8434 8438 8440
##      0      0      1      0      1      0      0      0      0      0      0      0      1      1      1      1
## 8446 8451 8453 8454 8462 8464 8469 8472 8473 8479 8481 8487 8491 8494 8499 8501
##      0      1      0      1      0      0      0      0      1      0      1      0      0      1      1      0
## 8504 8514 8521 8526
##      1      0      1      1
## Levels: 0 1

# Making the Confusion Matrix
cm = table(test_set[, 17], y_pred)
cm

##      y_pred
##      0      1
## 0 1092  148
## 1  144  748

print("=====Random Forest=====")

## [1] "=====Random Forest====="
```

```

library(ggplot2)
library(lattice)
library(caret)
confusionMatrix(cm)

## Confusion Matrix and Statistics
##
##      y_pred
##      0      1
## 0 1092  148
## 1  144  748
##
##              Accuracy : 0.863
##              95% CI : (0.8477, 0.8774)
##      No Information Rate : 0.5797
##      P-Value [Acc > NIR] : <2e-16
##
##              Kappa : 0.7188
##
##  Mcnemar's Test P-Value : 0.8606
##
##      Sensitivity : 0.8835
##      Specificity : 0.8348
##      Pos Pred Value : 0.8806
##      Neg Pred Value : 0.8386
##      Prevalence : 0.5797
##      Detection Rate : 0.5122
##      Detection Prevalence : 0.5816
##      Balanced Accuracy : 0.8592
##
##      'Positive' Class : 0
##

#=====
# Logistic Regression
rm(list=ls())
# Importing the dataset
dataset = read.csv('train.csv')

dataset$days_elapsed_old[dataset$days_elapsed_old<1] <- 0
dataset[ dataset == "na" ] <- NA

#Factor like columns
dataset$job=as.integer(as.factor(dataset$job))
dataset$marital=as.integer(as.factor(dataset$marital))
dataset$education=as.integer(as.factor(dataset$education))
dataset$device=as.integer(as.factor(dataset$device))
dataset$outcome_old=as.integer(as.factor(dataset$outcome_old))
dataset[is.na(dataset)] <- 0

# Encoding the target feature as factor

# Splitting the dataset into the Training set and Test set
# install.packages('caTools')

```

```

library(caTools)
set.seed(123)
split = sample.split(dataset$y, SplitRatio = 0.75)
training_set = subset(dataset, split == TRUE)
test_set = subset(dataset, split == FALSE)

# Feature Scaling
training_set[,1:16] = scale(training_set[,1:16])
test_set[-17] = scale(test_set[-17]) #removes third column alone

#fitting logistic regression to the training set
classifier = glm(formula = y ~ .,
                 family = binomial, #for logistic reg mention binomial
                 data = training_set)

#predicting the test set results
prob_pred = predict(classifier, type = 'response', newdata = test_set[-17]) #use type = response for logi.
prob_pred #that will give the prob listed in t

```

##	2	5	6	7	11	15
##	0.8427212359	0.8821662255	0.9028108432	0.0186598106	0.2169430278	0.3187913640
##	16	18	20	21	31	36
##	0.4782303985	0.0252262726	0.8918778421	0.0823708773	0.5398358529	0.9370782562
##	37	41	42	43	56	59
##	0.8913892844	0.4664663820	0.1313229621	0.9997663729	0.5743939009	0.3115089978
##	63	64	67	69	83	86
##	0.7937269908	0.0859625570	0.4087884255	0.1425132971	0.0422665782	0.4473865664
##	89	101	110	115	124	125
##	0.2757521457	0.9812313577	0.1740192571	0.0947024431	0.7163243271	0.4165784021
##	128	140	143	148	149	151
##	0.4296399412	0.0335868187	0.0319389717	0.1773857982	0.5745044887	0.3250740408
##	154	155	158	160	163	164
##	0.7786082651	0.3085817106	0.9695318767	0.0662565336	0.5687672687	0.4890073552
##	165	167	169	177	179	182
##	0.2853670543	0.4348449232	0.6442354423	0.0745776835	0.6486269487	0.1857915133
##	183	187	194	197	201	202
##	0.0708530589	0.1128402316	0.0491328778	0.0802314221	0.1471116990	0.3279233390
##	205	211	217	218	220	225
##	0.0334192875	0.8710546893	0.6143187140	0.9413090392	0.0947044261	0.3578358306
##	226	229	236	245	247	252
##	0.0641709178	0.0363876683	0.0775469253	0.2908783168	0.0921471272	0.1170099652
##	254	256	261	264	267	270
##	0.4497986798	0.2034595580	0.4151035300	0.0190968460	0.0240391340	0.9215563484
##	274	279	283	290	291	293
##	0.0759669296	0.9394466496	0.0393926463	0.0361543395	0.3344552201	0.0648951230
##	298	301	303	314	321	322
##	0.1261714249	0.1087129081	0.9863396872	0.9277931228	0.8939585959	0.4903711819
##	323	324	326	334	339	341
##	0.4928253065	0.7112224833	0.2807707676	0.4819204009	0.4355484823	0.1071816304
##	346	349	350	351	353	365
##	0.2636266654	0.1807861423	0.0884365856	0.1168121487	0.0544154271	0.0674707529
##	366	367	371	374	383	385
##	0.2522105678	0.0269585568	0.2242389735	0.1719779149	0.4713247761	0.3476375054
##	389	397	399	404	405	406

##	0.7266221212	0.0747423394	0.9435962397	0.2522363460	0.5452210213	0.9728591354
##	407	413	414	420	421	422
##	0.0288847232	0.0305090071	0.7126591138	0.5318296010	0.9504087249	0.1978523459
##	438	439	445	446	449	450
##	0.9997068637	0.7108999840	0.0257316855	0.0250372955	0.9047651061	0.0199325408
##	456	464	468	474	476	481
##	0.8836093246	0.5124794466	0.9560729935	0.9956181875	0.1860956286	0.5639502105
##	485	499	500	506	509	512
##	0.5851835588	0.1785487766	0.6310813306	0.9229876968	0.9985369873	0.1843733740
##	513	520	524	528	531	532
##	0.6237993704	0.3901467037	0.0561138441	0.0658393600	0.9938732259	0.2421568121
##	535	537	540	551	552	554
##	0.1477377347	0.0487187701	0.7761498993	0.9558383406	0.1288449874	0.6868315254
##	555	559	560	570	573	574
##	0.1430898977	0.3054583838	0.2362215718	0.0355941389	0.9909474463	0.9560547912
##	587	589	598	599	603	604
##	0.1817102042	0.1554311241	0.7838649810	0.6184195435	0.4718718522	0.1898879173
##	605	607	608	616	622	626
##	0.1146138127	0.9925906279	0.5300779790	0.9907058228	0.7309126164	0.5722619669
##	633	634	637	638	649	654
##	0.1408050483	0.3626958523	0.5883584610	0.3926272572	0.1951526293	0.1938996218
##	655	656	657	662	666	667
##	0.0921989075	0.0292796161	0.3893592674	0.1298407846	0.7935512233	0.4043473978
##	668	669	673	674	675	676
##	0.6969627490	0.9889617743	0.1029577406	0.4891231419	0.2064024618	0.8422276778
##	682	683	685	689	692	705
##	0.8304146710	0.1806566008	0.9564386029	0.1178877679	0.4013439847	0.0607249397
##	711	717	719	721	725	726
##	0.3550598446	0.0464215041	0.6235766711	0.9232803431	0.2141630071	0.7083231863
##	727	729	731	734	737	741
##	0.2869430641	0.7155059260	0.3685152170	0.1806942522	0.1717116056	0.2229360731
##	745	751	756	762	766	768
##	0.6920740792	0.4424922894	0.5680989909	0.2619537970	0.0228702374	0.2458887409
##	774	775	778	780	784	788
##	0.5016639889	0.7452189692	0.5410156219	0.7814342533	0.1609506864	0.0215661595
##	794	796	797	802	803	808
##	0.4271086104	0.0222720713	0.7556824105	0.1915191897	0.0348983153	0.2146697029
##	816	818	821	822	827	829
##	0.9589687551	0.5050111294	0.1112132641	0.0767809496	0.0365161363	0.0524727299
##	832	838	842	848	849	860
##	0.8748462750	0.4699182576	0.8891908062	0.0812118239	0.0735926433	0.4656230432
##	861	865	870	876	878	880
##	0.0557566068	0.7051716792	0.9318625904	0.1060984608	0.1314967294	0.2905341802
##	881	887	888	892	893	900
##	0.2360140678	0.1477326107	0.5947489296	0.9764178255	0.1446939211	0.4160096997
##	901	907	913	915	919	922
##	0.8641304264	0.0617085341	0.0659973686	0.3648946791	0.8629790976	0.0910740390
##	923	925	927	931	936	944
##	0.1116908111	0.1253703961	0.2948226402	0.4092255974	0.7767598834	0.3765226691
##	951	953	954	958	959	960
##	0.4914799935	0.1091224283	0.0103772116	0.8755572843	0.3555408609	0.9324593134
##	962	964	967	971	982	998
##	0.0361181183	0.7103004474	0.9999892499	0.1441943894	0.1160339250	0.3229467797
##	999	1003	1008	1013	1026	1028

##	0.4253276635	0.7674064655	0.0088476985	0.8373842235	0.4253198204	0.1660452531
##	1034	1042	1047	1049	1050	1056
##	0.4097443909	0.1647724817	0.0727836466	0.5366887926	0.6808192929	0.9222773902
##	1062	1069	1070	1081	1085	1089
##	0.1244843092	0.0271962991	0.0960874278	0.0277840702	0.1325917066	0.3064933411
##	1093	1096	1111	1119	1129	1131
##	0.9392259990	0.9110331073	0.7072247691	0.0642078220	0.9523021765	0.3692241392
##	1133	1134	1138	1140	1142	1146
##	0.4726987178	0.3928184806	0.3995160613	0.6914438658	0.2972268619	0.2203297890
##	1147	1148	1153	1156	1162	1167
##	0.2485347265	0.1984151885	0.2378029833	0.2901130859	0.9780776329	0.4702735926
##	1170	1175	1176	1182	1185	1193
##	0.2894466912	0.2219644016	0.1234249490	0.1447607725	0.1270059839	0.4976027641
##	1198	1202	1213	1214	1225	1232
##	0.3797609459	0.2684518375	0.7713773280	0.6661727088	0.2170191551	0.9549262628
##	1240	1241	1244	1249	1252	1257
##	0.0426274297	0.4497419714	0.0914249967	0.1525578844	0.8166379190	0.8279353267
##	1262	1266	1269	1270	1278	1284
##	0.1604035266	0.4203999228	0.3223071987	0.6230778123	0.2486835293	0.1432506747
##	1287	1290	1294	1303	1305	1313
##	0.4368702189	0.9758617997	0.3779648809	0.2073209566	0.7266935901	0.6969286325
##	1318	1319	1329	1330	1335	1338
##	0.9916972267	0.1592206642	0.6030155395	0.1804460379	0.3319056535	0.9202467975
##	1339	1341	1344	1347	1349	1356
##	0.2286854102	0.8993638911	0.4250237499	0.2376842729	0.9539266428	0.1958100338
##	1357	1364	1367	1368	1373	1374
##	0.2896726713	0.0735401322	0.8538813145	0.6491633640	0.1184148319	0.0268417395
##	1375	1379	1383	1385	1389	1396
##	0.1781229313	0.6191771769	0.6563237414	0.3691258607	0.7112660397	0.5852901053
##	1404	1406	1407	1408	1410	1412
##	0.0312643094	0.3496367143	0.2178910196	0.5945048427	0.0454531346	0.7925047080
##	1419	1421	1423	1427	1428	1429
##	0.1076481738	0.7110853030	0.0639043123	0.8973427476	0.2843182283	0.9777055865
##	1432	1436	1443	1446	1451	1453
##	0.2633429186	0.6591995649	0.2026881767	0.1628088389	0.8940015333	0.6836270582
##	1457	1463	1466	1473	1475	1476
##	0.0933092172	0.0945007290	0.9994375248	0.0100189687	0.5725840576	0.9534935245
##	1477	1478	1479	1481	1483	1484
##	0.0662934419	0.3187350804	0.1349942921	0.7604297498	0.9998424290	0.5132665654
##	1485	1487	1492	1496	1499	1501
##	0.2143602871	0.0711317231	0.9980858769	0.0726554649	0.2446550204	0.9306798326
##	1507	1508	1512	1516	1520	1523
##	0.4695217542	0.0236461989	0.5313207212	0.9952790696	0.9944905381	0.8991978325
##	1527	1529	1533	1546	1547	1548
##	0.9749355245	0.0338309017	0.7134616472	0.1273410510	0.0978594805	0.6629413774
##	1552	1553	1558	1561	1565	1569
##	0.1321637715	0.1649268081	0.0659345068	0.9961129412	0.4656292063	0.8686744873
##	1571	1577	1580	1581	1599	1619
##	0.4439755099	0.3628248722	0.8225579230	0.7438543505	0.2542423470	0.0859160783
##	1621	1623	1634	1638	1640	1641
##	0.5863233115	0.2290719033	0.9968294764	0.3487307984	0.0534887697	0.4055182734
##	1645	1652	1665	1673	1677	1683
##	0.3992785358	0.0842061275	0.6315174766	0.2124532927	0.7514333976	0.1829086112
##	1689	1694	1718	1719	1723	1727



##	0.1443069466	0.0394893114	0.9968683082	0.2222654940	0.1891133811	0.2879375404
##	1729	1730	1731	1735	1736	1737
##	0.8942481348	0.3538740948	0.0443485808	0.6107358846	0.1386613420	0.5322744141
##	1740	1743	1746	1751	1752	1754
##	0.4130585392	0.0612076379	0.5401391171	0.7133578530	0.0513424429	0.2088307685
##	1760	1761	1763	1767	1769	1782
##	0.0943339904	0.9132528622	0.2293055812	0.7459048583	0.2475140695	0.2213419933
##	1783	1792	1796	1803	1806	1807
##	0.1630408037	0.4768674687	0.4698006257	0.1038980833	0.9400645588	0.8207205037
##	1808	1816	1821	1823	1824	1827
##	0.5087640920	0.5267291255	0.0724925045	0.4427365125	0.3800695806	0.0826483692
##	1830	1837	1839	1840	1843	1846
##	0.2480974950	0.0419572027	0.5508859607	0.3225809292	0.2323319834	0.1484657975
##	1849	1855	1856	1858	1864	1866
##	0.9174203904	0.2490171509	0.3840053654	0.4301348560	0.0677142983	0.2291179660
##	1870	1872	1876	1878	1882	1891
##	0.3681074032	0.9123854963	0.8064131735	0.0304309181	0.8115530552	0.2711860728
##	1893	1900	1904	1925	1926	1928
##	0.8930156631	0.8114115490	0.8817794229	0.1339574603	0.6003570547	0.6825702730
##	1929	1933	1939	1943	1956	1964
##	0.6451832272	0.9503656656	0.6245100653	0.5224918914	0.5576304422	0.2280296923
##	1966	1972	1975	1976	1981	1991
##	0.0358572495	0.5871142200	0.4211996623	0.0501034202	0.2754341083	0.4527744566
##	1999	2003	2004	2007	2008	2015
##	0.0401270557	0.4685530412	0.0194887838	0.5071606176	0.0222849321	0.0727350251
##	2017	2018	2020	2027	2028	2029
##	0.1926296044	0.0863218798	0.6082979470	0.2434299947	0.9627867061	0.1136993008
##	2035	2042	2046	2049	2050	2052
##	0.8086635572	0.0154586337	0.6276217442	0.1412160369	0.0979448521	0.0272088834
##	2053	2056	2073	2074	2076	2081
##	0.1479069796	0.2650065141	0.0822077558	0.0709180660	0.9100098313	0.4298276841
##	2083	2084	2088	2093	2098	2102
##	0.9019560533	0.0242904176	0.2427731595	0.1396748569	0.0289359568	0.1520883438
##	2107	2113	2115	2117	2120	2134
##	0.8285441382	0.5602458251	0.0422436341	0.2995303126	0.0934076216	0.0776665766
##	2135	2139	2140	2145	2150	2152
##	0.8966785017	0.2914481999	0.9851601230	0.8528181831	0.2266910292	0.0272487060
##	2154	2155	2162	2167	2168	2174
##	0.4489733428	0.7601137160	0.0687093613	0.7508212237	0.0506460273	0.8957225793
##	2175	2188	2198	2199	2203	2216
##	0.6952050234	0.1806210718	0.2485281308	0.3381455444	0.2662131669	0.4777250853
##	2217	2218	2222	2226	2230	2235
##	0.5395849862	0.2546567994	0.1286297806	0.4938980711	0.7966397403	0.0641362168
##	2241	2247	2253	2254	2255	2256
##	0.6081628295	0.2743655682	0.7125205463	0.8052577672	0.0950199640	0.3102040875
##	2259	2264	2270	2272	2275	2277
##	0.7561643976	0.6038453448	0.2608721981	0.8243596385	0.1876170201	0.0502255464
##	2278	2282	2283	2289	2309	2315
##	0.2927905966	0.0584992730	0.3293507268	0.1067857007	0.8675376222	0.0615541029
##	2316	2322	2327	2329	2330	2332
##	0.9053783584	0.5192514381	0.0775227491	0.3392253329	0.3115180999	0.0957362523
##	2338	2345	2349	2353	2356	2357
##	0.3516060132	0.9879875884	0.0273683210	0.0822066435	0.8389277296	0.3099140395
##	2365	2373	2376	2377	2385	2389

##	0.2224820690	0.0028610951	0.3369252831	0.0917237933	0.9819106388	0.2099796521
##	2391	2392	2394	2400	2403	2406
##	0.8439687008	0.5553824519	0.0937385325	0.0708382102	0.0940618298	0.2408779424
##	2408	2412	2421	2426	2430	2431
##	0.6911267050	0.0988565615	0.5400341764	0.0518749851	0.4292523328	0.4754673027
##	2432	2436	2439	2447	2457	2460
##	0.8413828114	0.8508754431	0.3200916623	0.0516918664	0.9194538571	0.7791254208
##	2462	2464	2465	2466	2469	2474
##	0.2123976274	0.2525073137	0.1513729360	0.1854977668	0.1859640130	0.0134912665
##	2475	2479	2486	2488	2491	2504
##	0.4181681446	0.9577123014	0.9959496987	0.1217473607	0.4522092756	0.1234472670
##	2505	2510	2514	2515	2516	2520
##	0.1769548853	0.7550666412	0.2230115425	0.9784893996	0.1556974514	0.2169290156
##	2521	2523	2526	2531	2534	2535
##	0.5860864120	0.0731200742	0.0236594116	0.1500207504	0.8727439548	0.9502002077
##	2537	2538	2541	2542	2546	2554
##	0.0735093660	0.5753222426	0.0537980214	0.3285734914	0.1984301975	0.1470841758
##	2563	2574	2579	2580	2581	2582
##	0.7318241949	0.9486717821	0.9504032611	0.3392017115	0.0244135958	0.2295738655
##	2586	2591	2594	2595	2598	2601
##	0.4236747566	0.3827888793	0.9212544573	0.9881492377	0.4269532214	0.2590609454
##	2610	2615	2619	2626	2627	2631
##	0.0716685989	0.2525619143	0.3992998690	0.0085681920	0.0312573333	0.1107583299
##	2632	2639	2640	2644	2645	2647
##	0.8562248544	0.0228084708	0.0871187328	0.2728003155	0.9393093219	0.0582564607
##	2649	2656	2657	2658	2659	2661
##	0.6856478355	0.2969946438	0.9886735129	0.1262449353	0.3296186132	0.0746366892
##	2662	2669	2672	2675	2679	2688
##	0.4905002419	0.1852335816	0.0662754215	0.9096799402	0.2067487864	0.1897091436
##	2693	2695	2696	2699	2703	2704
##	0.1085576624	0.2528134131	0.4093569565	0.8618764992	0.8409078349	0.5463065118
##	2712	2722	2736	2744	2746	2747
##	0.4197600909	0.3744528275	0.2435908490	0.6592382662	0.8906925080	0.9837748701
##	2751	2753	2754	2755	2758	2763
##	0.0328041521	0.3357268798	0.1575353229	0.9993650028	0.2963876886	0.0740177496
##	2767	2768	2769	2770	2771	2780
##	0.2076164351	0.8406993826	0.0554582986	0.0919165658	0.0425595064	0.6657924411
##	2784	2785	2787	2793	2798	2803
##	0.1187566629	0.5884104387	0.0530951677	0.0948859038	0.8953870671	0.5505200104
##	2804	2807	2810	2812	2820	2825
##	0.6785942643	0.0401199205	0.3798595000	0.1182444526	0.0569809456	0.0232915532
##	2832	2834	2839	2843	2848	2850
##	0.0685381114	0.4882129302	0.1310920672	0.4897079098	0.0664784593	0.7266160145
##	2856	2866	2872	2873	2878	2879
##	0.0807471388	0.0399530629	0.1603880617	0.2348782158	0.8555957539	0.8307386546
##	2883	2892	2899	2900	2902	2905
##	0.4959174261	0.4789854385	0.8348903245	0.2516198999	0.2562131258	0.8934900633
##	2906	2915	2916	2923	2927	2934
##	0.2143640274	0.8686909723	0.5354828604	0.4183308875	0.9271098187	0.7310658442
##	2936	2940	2943	2945	2949	2951
##	0.0138530254	0.7183457748	0.9587591670	0.3883120194	0.0514534096	0.5728270599
##	2957	2959	2962	2963	2965	2967
##	0.3302364489	0.3071722666	0.9197506505	0.3551373043	0.1733677362	0.3346978440
##	2981	2984	2985	2989	3004	3011

##	0.5893412667	0.1394203416	0.1753264450	0.4744124471	0.6311608735	0.0908300876
##	3017	3028	3037	3042	3044	3050
##	0.7623452343	0.9313438968	0.8457610970	0.0535285187	0.8335025441	0.2272082406
##	3058	3059	3061	3064	3067	3068
##	0.1981258950	0.0557004971	0.2788578199	0.0774541850	0.1431671221	0.5990807697
##	3071	3074	3075	3076	3082	3089
##	0.5416465378	0.2494667539	0.1160683618	0.2392721122	0.9602626701	0.0612276868
##	3093	3095	3098	3101	3102	3105
##	0.1727674829	0.0219640226	0.3351571968	0.3213291192	0.8679527541	0.3401627658
##	3106	3111	3119	3127	3134	3136
##	0.5639532595	0.2240443390	0.9926235035	0.2897832696	0.4689467574	0.2537888393
##	3137	3147	3151	3160	3172	3173
##	0.9426323927	0.3731398426	0.6994971159	0.3333262218	0.8482817460	0.0413185870
##	3180	3181	3185	3186	3189	3193
##	0.0623718474	0.7769473994	0.3361783893	0.1077812142	0.0461094719	0.0235834118
##	3196	3198	3204	3206	3210	3212
##	0.0961695384	0.5577562773	0.9879483370	0.2106631918	0.0119392229	0.4049746425
##	3217	3227	3231	3233	3238	3240
##	0.3014503211	0.0945538469	0.1788508616	0.2909468692	0.7767541440	0.9079793509
##	3243	3245	3246	3249	3254	3257
##	0.0636548451	0.0481725984	0.1020985465	0.2060419259	0.0781750946	0.1506296300
##	3259	3265	3272	3282	3285	3289
##	0.9941212502	0.1523289468	0.3668281833	0.2153783366	0.5523311242	0.3457874938
##	3294	3305	3310	3323	3325	3328
##	0.9250561286	0.3191675595	0.1995700605	0.1847409032	0.0415710905	0.1275260573
##	3329	3330	3339	3341	3344	3348
##	0.4353054681	0.5448268066	0.6067314503	0.0561520124	0.1288310395	0.1183128057
##	3350	3354	3356	3357	3359	3363
##	0.8419189361	0.1129435667	0.0937504951	0.8801542729	0.2708397972	0.0778580599
##	3364	3374	3384	3387	3395	3399
##	0.9114077437	0.6731301265	0.4034856566	0.0840248946	0.0525002735	0.0912802142
##	3402	3405	3408	3409	3416	3417
##	0.1260280362	0.6560899420	0.3330102721	0.2710499925	0.2249028230	0.9997676671
##	3419	3427	3430	3431	3432	3434
##	0.2100437483	0.6260818847	0.0224688527	0.2814230395	0.2375014894	0.1436717495
##	3435	3438	3441	3450	3461	3462
##	0.1332316352	0.7122450506	0.8264234762	0.2848711974	0.6064780128	0.8579630297
##	3464	3466	3467	3477	3478	3483
##	0.9764807292	0.8546973008	0.8883855654	0.7354695581	0.4173856592	0.9755385857
##	3490	3493	3499	3503	3506	3510
##	0.2007675159	0.0900084534	0.0787535593	0.9699169854	0.4604403677	0.0498505265
##	3514	3523	3524	3533	3534	3541
##	0.0401349163	0.2484859628	0.1702881374	0.9564778650	0.1732860368	0.8152836064
##	3542	3544	3545	3548	3549	3551
##	0.0928077225	0.1054540146	0.8467342428	0.1822929661	0.9289419777	0.9449086954
##	3558	3559	3563	3579	3580	3583
##	0.9530347406	0.3996338726	0.0578139013	0.5408528684	0.8661936732	0.4714776534
##	3585	3587	3590	3591	3592	3596
##	0.7774036045	0.9425662432	0.7057402240	0.3227021639	0.6439194077	0.9031712273
##	3597	3598	3602	3603	3608	3610
##	0.1809413054	0.0156853484	0.6000962204	0.9635602512	0.9233613173	0.8112507070
##	3613	3620	3621	3624	3626	3630
##	0.9299077190	0.3368887782	0.0987129184	0.2046181633	0.2396969175	0.8055905685
##	3631	3641	3646	3647	3650	3654

##	0.0598637532	0.0359034612	0.0942105627	0.0306617740	0.1625046267	0.4641136741
##	3659	3660	3664	3676	3689	3692
##	0.8568937067	0.8121691560	0.1062234711	0.3971145891	0.8840727644	0.0420800474
##	3695	3700	3706	3707	3711	3714
##	0.8912217465	0.4270515317	0.5848266544	0.8447032929	0.8143755338	0.3317972706
##	3715	3724	3726	3727	3741	3742
##	0.5545959967	0.1069680011	0.9632694444	0.9601447642	0.4714199207	0.0525880494
##	3749	3755	3756	3758	3761	3763
##	0.0597226188	0.9724439668	0.3011939325	0.0469393822	0.2242203479	0.2878899793
##	3766	3775	3778	3779	3781	3782
##	0.8453655906	0.4361682628	0.7747031230	0.5352676126	0.9954185032	0.7019850442
##	3789	3795	3801	3808	3819	3823
##	0.0490998083	0.9056396441	0.4318038313	0.0220423738	0.0350551774	0.1150542879
##	3824	3833	3836	3839	3841	3846
##	0.6968603749	0.0946175663	0.0364749031	0.9646896379	0.2444582977	0.9050725699
##	3847	3851	3872	3877	3878	3881
##	0.2565790497	0.4168521304	0.8338718922	0.8824782089	0.9262008790	0.4444298622
##	3882	3883	3893	3895	3896	3898
##	0.8527917356	0.1337330351	0.4570376068	0.1486961878	0.6614409292	0.0589390356
##	3899	3900	3906	3914	3917	3924
##	0.1463800721	0.7531700293	0.0662393534	0.0303803573	0.1917074021	0.7375651519
##	3927	3928	3937	3938	3941	3949
##	0.1217837024	0.6075292391	0.9630803599	0.2392723781	0.1158625361	0.0280472580
##	3951	3952	3958	3960	3962	3963
##	0.8853412896	0.3544323381	0.3522116355	0.3613667539	0.0018757364	0.8375796807
##	3965	3967	3969	3970	3972	3973
##	0.4871592082	0.1142163430	0.3174821881	0.3494369608	0.7365097147	0.9010331184
##	3974	3981	3982	3993	3996	3997
##	0.2492172174	0.4857407348	0.0342498551	0.1462963401	0.2350821213	0.0149845348
##	4002	4004	4006	4007	4015	4017
##	0.9500916039	0.0541247325	0.9512567014	0.3563767821	0.3803485130	0.1129220120
##	4018	4021	4033	4034	4047	4048
##	0.8183367456	0.4350403044	0.3824755514	0.1392874069	0.1341095471	0.1405059551
##	4056	4060	4065	4066	4069	4075
##	0.2341726994	0.0414591257	0.2041658491	0.0267700933	0.2673546834	0.1731993548
##	4076	4077	4078	4083	4088	4090
##	0.1390372218	0.1012559347	0.5856361423	0.3636563509	0.1309853199	0.5702965507
##	4092	4095	4097	4100	4108	4109
##	0.4292284790	0.0321404067	0.1316028998	0.3155386504	0.3233328261	0.0845977365
##	4110	4112	4113	4122	4127	4131
##	0.8901201459	0.6497612511	0.3500260650	0.0450194133	0.9828685386	0.4670257089
##	4134	4135	4136	4138	4139	4140
##	0.4298620912	0.3856463463	0.8413777842	0.2794919091	0.4124368260	0.1012918358
##	4150	4151	4152	4157	4158	4162
##	0.3277566865	0.1213190493	0.1087050764	0.1777857165	0.1778009421	0.0783687445
##	4166	4167	4169	4171	4175	4178
##	0.4961566627	0.2092655522	0.9369991583	0.4357908888	0.9853278773	0.5635174516
##	4184	4185	4190	4191	4197	4204
##	0.0420461089	0.1235911568	0.1494940634	0.8342896677	0.0843536946	0.1289334776
##	4205	4208	4209	4210	4212	4215
##	0.0496792137	0.3443435574	0.4961030476	0.1336745932	0.3995040216	0.0380265385
##	4217	4218	4219	4225	4236	4238
##	0.1825957205	0.9376252971	0.9477572037	0.9999988732	0.9538037599	0.9691496631
##	4241	4243	4245	4247	4248	4249

##	0.8893724250	0.2954437341	0.9995124705	0.5102670033	0.1792658701	0.2409873504
##	4256	4259	4266	4269	4271	4280
##	0.0745432720	0.7911113361	0.4378311746	0.0284453578	0.8422923281	0.6969160803
##	4282	4283	4287	4290	4291	4293
##	0.5213929527	0.2717748863	0.0994264630	0.9987522094	0.1944170081	0.5890457113
##	4317	4326	4329	4333	4334	4337
##	0.3741532225	0.3015874805	0.0606849925	0.2088872423	0.0288378170	0.1851658904
##	4340	4345	4353	4359	4364	4366
##	0.2722593476	0.3049130176	0.6651766066	0.2196373221	0.0177696033	0.9999856334
##	4371	4374	4377	4378	4379	4381
##	0.0122132245	0.9694009242	0.1555391763	0.0406503762	0.9993492521	0.1015759512
##	4388	4390	4391	4396	4403	4407
##	0.6813052661	0.1290921899	0.4023566802	0.1733170522	0.2165043815	0.2947766029
##	4409	4413	4415	4417	4419	4423
##	0.9392025826	0.6053043161	0.3417191846	0.1706744621	0.9771299277	0.8440421210
##	4424	4427	4428	4431	4433	4438
##	0.7824580397	0.1633938603	0.3665062151	0.1560267550	0.0451435184	0.1182113415
##	4439	4440	4444	4445	4448	4456
##	0.0717282971	0.1953724172	0.7063857485	0.9066637858	0.4724090414	0.0713963731
##	4461	4464	4469	4479	4481	4482
##	0.9091448464	0.0525093524	0.3159181805	0.2860336081	0.1641413962	0.3198218788
##	4484	4485	4486	4489	4492	4498
##	0.9381447196	0.0292182592	0.0523313449	0.0662240143	0.3030051284	0.4086732964
##	4502	4505	4507	4509	4510	4512
##	0.4783155496	0.0409915903	0.0667369320	0.1416218515	0.8742573280	0.3107851229
##	4517	4520	4523	4524	4526	4533
##	0.9974063951	0.3703691030	0.8883882481	0.1346067632	0.1663101658	0.0915806856
##	4534	4540	4542	4548	4552	4560
##	0.7502477235	0.6038673221	0.7730068392	0.3199207891	0.1215470781	0.5558587330
##	4572	4578	4582	4585	4596	4625
##	0.5967526913	0.7936461397	0.9719128228	0.3626259653	0.7854207311	0.0226809740
##	4630	4633	4642	4654	4656	4670
##	0.8582837645	0.3185600082	0.8830764791	0.2901082359	0.2967615247	0.0973109569
##	4674	4678	4688	4693	4698	4703
##	0.6837848935	0.8432161794	0.8334425651	0.0781396834	0.9389816748	0.9978070368
##	4715	4716	4722	4723	4731	4739
##	0.0731227022	0.7794622587	0.1982648156	0.5841394929	0.3101475343	0.2266164383
##	4742	4746	4748	4753	4757	4759
##	0.9555105123	0.6248732064	0.5823512200	0.5365492341	0.1711789640	0.4544200583
##	4763	4771	4775	4778	4780	4783
##	0.2010082792	0.6102897037	0.4387844948	0.1654514149	0.1258391674	0.2126878797
##	4784	4786	4787	4788	4801	4806
##	0.9670247192	0.0800784538	0.2338408794	0.0775180306	0.6217834422	0.9309769420
##	4807	4811	4817	4826	4831	4840
##	0.2439174512	0.6157479418	0.4371762049	0.2608287623	0.2326006305	0.8241018340
##	4845	4854	4858	4863	4868	4870
##	0.6906224342	0.1588326569	0.9807735606	0.6392213550	0.0492253736	0.4611767291
##	4871	4877	4879	4889	4890	4891
##	0.6129428209	0.5896720145	0.1615290378	0.3915460629	0.1057443593	0.1777446848
##	4899	4902	4905	4906	4908	4913
##	0.9998396900	0.2778292200	0.3070768803	0.3707977483	0.3193128137	0.0815017980
##	4917	4919	4924	4925	4926	4927
##	0.0922103244	0.2143290319	0.7784782882	0.0939202462	0.3139402549	0.0454963259
##	4935	4936	4937	4938	4939	4940

##	0.0079395011	0.6370939859	0.2777620649	0.4875618415	0.1548195176	0.9761417394
##	4946	4947	4950	4959	4965	4966
##	0.2881037836	0.1181279051	0.2493424390	0.4046349872	0.2299220646	0.8178780062
##	4968	4970	4972	4978	4980	4984
##	0.2068794048	0.3317845888	0.4001730597	0.7117677264	0.2364541752	0.8774038752
##	4987	4988	4989	4991	5000	5003
##	0.1650018029	0.0594425474	0.6330110200	0.1593990270	0.2351575274	0.1563392808
##	5006	5009	5011	5014	5016	5017
##	0.1631741883	0.0380200113	0.9379977603	0.2011142950	0.8880658200	0.2370719719
##	5020	5021	5029	5039	5041	5051
##	0.9999490073	0.0809664365	0.9888466153	0.0648788522	0.2502088009	0.0553040465
##	5056	5059	5064	5070	5074	5075
##	0.7712436559	0.9525018009	0.0346193754	0.8476492540	0.8558507469	0.7916716967
##	5087	5089	5093	5095	5100	5101
##	0.1417330125	0.2291436262	0.2071424180	0.3775700726	0.6983452744	0.8845606935
##	5116	5123	5124	5127	5131	5133
##	0.8104054335	0.8180858344	0.4882710023	0.0414775075	0.7855138996	0.5293915853
##	5134	5137	5138	5140	5144	5153
##	0.0826812427	0.6434956578	0.5440350304	0.7886679922	0.0735338809	0.4635222148
##	5154	5163	5166	5172	5186	5188
##	0.8863427272	0.3770581114	0.3315216769	0.0867182807	0.9160718469	0.0392503521
##	5189	5193	5194	5197	5204	5206
##	0.2023397402	0.6010466752	0.2111862552	0.1160546640	0.0935222897	0.7369836053
##	5211	5214	5215	5216	5218	5226
##	0.9740774601	0.0505456983	0.0200509757	0.0599115749	0.9643174083	0.1017848448
##	5227	5230	5236	5244	5245	5248
##	0.9533737443	0.8935543970	0.3281964847	0.7662934387	0.4234872433	0.1350378206
##	5250	5251	5254	5261	5262	5264
##	0.2229800996	0.4917021812	0.2413669324	0.1031592507	0.6759426107	0.6234937448
##	5265	5268	5269	5273	5275	5276
##	0.9414893412	0.1456513789	0.3427045608	0.1152147524	0.2875529536	0.3051923815
##	5280	5283	5284	5294	5295	5296
##	0.6632414051	0.9653857149	0.9346929836	0.4981978529	0.4118956624	0.8510009851
##	5300	5303	5305	5310	5330	5335
##	0.2293930924	0.0229380202	0.6314888885	0.8843923611	0.0073008171	0.1064044041
##	5352	5353	5355	5357	5358	5359
##	0.0191091203	0.2772269222	0.8017274842	0.0493279297	0.5678595741	0.9913397491
##	5360	5363	5365	5366	5369	5371
##	0.1637793820	0.1068500649	0.7412021799	0.1719271453	0.1720576154	0.1361237639
##	5376	5378	5382	5385	5389	5393
##	0.1316850328	0.5673187511	0.3438157488	0.4361321401	0.9909400834	0.4316636446
##	5400	5402	5405	5406	5407	5410
##	0.7513054520	0.1440631952	0.9550271172	0.1996878920	0.8286866329	0.1455543626
##	5411	5417	5418	5419	5431	5432
##	0.2440419508	0.0692652698	0.5576201809	0.7484172947	0.0662251971	0.5275204794
##	5433	5442	5449	5452	5454	5455
##	0.8705595394	0.9653546518	0.1641657905	0.2016589056	0.9481644019	0.0601856002
##	5462	5466	5472	5473	5474	5483
##	0.7925642035	0.0986079127	0.8982539127	0.6298345333	0.9674084270	0.4808372920
##	5491	5495	5503	5508	5510	5511
##	0.1853057139	0.8901305720	0.5004726125	0.1290207360	0.8726360499	0.5202757953
##	5513	5515	5520	5524	5529	5539
##	0.0498392339	0.0308041601	0.9484663259	0.6402121183	0.7875999894	0.2195308723
##	5545	5550	5551	5554	5566	5584

##	0.0596494962	0.0983445653	0.1591324578	0.6620892682	0.3986204033	0.0090319210
##	5589	5599	5602	5604	5607	5608
##	0.3657876167	0.1603072196	0.9959962344	0.3824281895	0.8470823746	0.9095675936
##	5614	5616	5621	5622	5628	5638
##	0.3035668499	0.2246701658	0.3167686787	0.9160510623	0.0788755518	0.4842219228
##	5642	5651	5654	5662	5666	5675
##	0.8595467139	0.0941071135	0.2619326892	0.2380027262	0.0426382511	0.0976039446
##	5676	5691	5693	5704	5707	5709
##	0.9841528897	0.1124614101	0.8328841375	0.5122996502	0.0970374135	0.9819160700
##	5710	5715	5718	5720	5723	5726
##	0.1096591979	0.4929192813	0.0252030760	0.7699172349	0.1563505357	0.9810987540
##	5729	5732	5736	5739	5740	5745
##	0.4585605105	0.5664395408	0.1986277213	0.1318253491	0.2849459147	0.5038176318
##	5746	5751	5752	5757	5758	5759
##	0.9060147756	0.3026174369	0.6916912063	0.1560227384	0.6536874818	0.0272457001
##	5765	5770	5772	5777	5779	5780
##	0.0989942450	0.3914544908	0.9309691954	0.1586885608	0.5309921505	0.1846826242
##	5782	5797	5806	5810	5811	5814
##	0.2578355382	0.9732447830	0.1537003261	0.2918518995	0.8959532306	0.4264892039
##	5820	5828	5835	5849	5857	5862
##	0.1378817860	0.1228178948	0.6639213244	0.6305886094	0.9593788201	0.0428873220
##	5865	5867	5868	5871	5875	5878
##	0.2993424229	0.5239602136	0.1734658432	0.9300815772	0.2424716299	0.2792958835
##	5881	5882	5887	5893	5901	5904
##	0.4569576088	0.1430461595	0.4020832685	0.8539205327	0.0402830364	0.1915323077
##	5909	5911	5922	5923	5924	5925
##	0.1604089008	0.0338109359	0.0326276827	0.0017919050	0.9512286728	0.0722594846
##	5930	5931	5932	5934	5937	5941
##	0.2136774227	0.8856957162	0.9332830988	0.1201137182	0.6626551788	0.3606996585
##	5946	5949	5951	5953	5963	5973
##	0.9915768093	0.5409876487	0.9772078022	0.4832150471	0.0244964110	0.0511865982
##	5977	5978	5980	5984	5993	5995
##	0.1680903349	0.8924745402	0.9731706225	0.5891310126	0.1324158674	0.6636704305
##	5998	6000	6004	6008	6015	6016
##	0.5322730939	0.2920653840	0.2378149814	0.0381094670	0.6695388059	0.7982738384
##	6018	6030	6035	6044	6047	6049
##	0.5580779224	0.0621572149	0.3319712645	0.9326326524	0.4453703588	0.0038999565
##	6050	6052	6056	6057	6059	6063
##	0.7911144149	0.5107677762	0.0171563365	0.2218190615	0.4969995118	0.3730819017
##	6065	6066	6068	6069	6076	6096
##	0.3378421979	0.1676492517	0.0519055809	0.3475178125	0.4993513265	0.5610380060
##	6097	6099	6100	6103	6106	6109
##	0.2020955609	0.5360540001	0.8381384776	0.5978917880	0.8321811425	0.5334258518
##	6115	6116	6117	6119	6121	6125
##	0.0776338466	0.1115972219	0.1659276751	0.2201545231	0.0185363324	0.8936276194
##	6129	6142	6143	6146	6148	6158
##	0.8947526646	0.2000381556	0.0449770094	0.4899273713	0.0497853716	0.1791584686
##	6159	6167	6168	6169	6177	6185
##	0.9618058653	0.3549463340	0.1689864571	0.7171206696	0.3822016750	0.9747337265
##	6190	6193	6196	6199	6200	6206
##	0.5117456149	0.5345687581	0.2805568139	0.6357565949	0.6154613986	0.9059392007
##	6211	6219	6221	6222	6223	6224
##	0.1975068649	0.9597540987	0.8407475901	0.1206905493	0.5681185945	0.2715846224
##	6228	6235	6237	6242	6246	6250

##	0.2656286487	0.1715638734	0.6210097672	0.0206886534	0.2713821656	0.0315542813
##	6258	6264	6265	6268	6272	6275
##	0.0850517128	0.2569796651	0.1268491103	0.0417794271	0.8843360209	0.2104757556
##	6284	6285	6290	6293	6294	6296
##	0.4248007505	0.4429775364	0.1307723055	0.0958076620	0.3469932174	0.2944125834
##	6306	6317	6324	6325	6328	6334
##	0.9614601451	0.1042474296	0.9645641673	0.7623110548	0.2040538364	0.3070560153
##	6337	6340	6344	6348	6350	6352
##	0.2225916747	0.2392021840	0.0033041793	0.1498003868	0.4330713602	0.1264324690
##	6353	6354	6358	6360	6361	6362
##	0.2934996954	0.8684607776	0.2503956603	0.8177809220	0.0231529730	0.8798898619
##	6363	6379	6380	6391	6393	6397
##	0.5860313756	0.2760761287	0.1361387624	0.3115374577	0.0508197580	0.6363852972
##	6400	6411	6415	6418	6419	6422
##	0.4551267958	0.9708876382	0.4221543476	0.2508128883	0.1840297345	0.4967950799
##	6423	6430	6437	6440	6444	6451
##	0.9390255872	0.2469377378	0.6463982496	0.4377681670	0.3370168231	0.5179222936
##	6460	6462	6463	6465	6466	6467
##	0.0273777069	0.1001022998	0.0204723756	0.5691295353	0.0500110292	0.6059019547
##	6469	6471	6477	6478	6481	6483
##	0.9432541608	0.2881662452	0.9079357698	0.0229290844	0.9177695999	0.2039581787
##	6485	6486	6487	6488	6489	6490
##	0.0490907065	0.0246059729	0.9983559968	0.0912424927	0.0265131983	0.1204492218
##	6491	6492	6497	6499	6502	6508
##	0.3270515323	0.1396073656	0.0474876002	0.1939831518	0.6780246347	0.1790118827
##	6515	6521	6522	6527	6532	6534
##	0.5173334940	0.6429027788	0.0708476650	0.9324213454	0.5946933828	0.0822442796
##	6535	6538	6549	6551	6553	6555
##	0.4766924316	0.6050499419	0.9677271937	0.0748165543	0.5895881206	0.3938677447
##	6559	6561	6563	6565	6566	6569
##	0.3222477819	0.9501677747	0.2035495722	0.9264531584	0.2895063933	0.4661301330
##	6575	6580	6582	6586	6614	6615
##	0.2824156580	0.8362934989	0.8615650168	0.1172630087	0.4068899470	0.6065829045
##	6617	6618	6620	6632	6646	6654
##	0.0821475307	0.3351175201	0.4067171621	0.0508315050	0.3138171245	0.1626492031
##	6658	6661	6663	6664	6665	6673
##	0.1508890623	0.1761548909	0.3134423932	0.8031785966	0.4640561162	0.0980603991
##	6677	6681	6682	6686	6687	6690
##	0.0825642143	0.8227168676	0.5939480955	0.5123273551	0.1687204679	0.6969819270
##	6693	6694	6695	6697	6698	6702
##	0.6670526336	0.3716476585	0.7361070328	0.6636851035	0.0710496947	0.1003426265
##	6705	6717	6724	6727	6728	6730
##	0.9595189831	0.8723037830	0.6300202336	0.0078000850	0.2970316902	0.7050377265
##	6732	6733	6738	6740	6743	6751
##	0.9888168305	0.2736542739	0.9877644076	0.4896425017	0.9389490055	0.1562002255
##	6756	6757	6760	6763	6767	6770
##	0.3172600729	0.0408282631	0.6099494678	0.8208625751	0.0572388170	0.0469243319
##	6771	6777	6782	6783	6784	6788
##	0.0380327749	0.8804977662	0.1031505485	0.3354008434	0.2456969363	0.2540935272
##	6790	6793	6797	6798	6803	6807
##	0.0460549694	0.6199676407	0.0656970219	0.0405781145	0.9193567393	0.0540881163
##	6813	6815	6825	6832	6837	6848
##	0.9487519657	0.1332157296	0.0488258895	0.5785278880	0.9589263561	0.0210270238
##	6850	6852	6855	6859	6864	6868



##	0.2538128427	0.0331246680	0.1551765829	0.0220730344	0.7793939296	0.0282473760
##	6871	6873	6885	6887	6891	6895
##	0.7496984225	0.9061561056	0.4444788976	0.0285568599	0.3926877320	0.9578586860
##	6900	6903	6909	6920	6922	6926
##	0.4827418702	0.4563419942	0.9496729034	0.2762986818	0.1529105968	0.3482972484
##	6930	6932	6941	6952	6953	6958
##	0.0455206511	0.2689672941	0.3439518771	0.4171572020	0.0602906824	0.6100458944
##	6961	6964	6974	6976	6977	6978
##	0.2718034761	0.1955415828	0.1053211143	0.2001991133	0.0401679076	0.0806522942
##	6979	6981	6986	6993	7002	7005
##	0.4376456407	0.9999985915	0.0554916172	0.9062567930	0.0690795954	0.9160852550
##	7015	7020	7022	7026	7028	7039
##	0.2345674649	0.8772176257	0.3888451538	0.4743550929	0.1471233025	0.3146291306
##	7043	7047	7058	7073	7074	7081
##	0.0181511339	0.3583186914	0.6454518834	0.2602659241	0.9826379471	0.3161787773
##	7083	7084	7085	7086	7087	7089
##	0.2757405495	0.0006488195	0.4130511215	0.5324344267	0.2539583922	0.9894953045
##	7094	7097	7098	7104	7110	7115
##	0.4176124354	0.1743528454	0.3463012718	0.0683841403	0.4933701363	0.9700878794
##	7118	7124	7129	7133	7138	7146
##	0.5328560793	0.0282403528	0.6041963066	0.8739758692	0.3452842674	0.0662225348
##	7155	7160	7164	7165	7169	7172
##	0.1303507373	0.1705661866	0.6416392054	0.1627737666	0.9017530777	0.0519841815
##	7173	7180	7187	7194	7196	7200
##	0.2818476219	0.5939577047	0.0292057501	0.2617446472	0.5755498632	0.0605195640
##	7202	7203	7204	7211	7215	7217
##	0.9893921661	0.0707037524	0.0276416881	0.0106452920	0.7469952659	0.1405979854
##	7220	7226	7237	7240	7241	7244
##	0.1426199013	0.2121144578	0.7304083883	0.9891814556	0.4592619060	0.0219609608
##	7248	7249	7254	7263	7264	7265
##	0.3687419188	0.5315868393	0.8449246971	0.8465078567	0.1125391138	0.1041187081
##	7267	7268	7271	7273	7274	7277
##	0.4393763170	0.1248833675	0.7041436878	0.2161027400	0.0182300999	0.2253392140
##	7278	7283	7289	7295	7298	7300
##	0.9158116415	0.1953998257	0.1467724007	0.3061117763	0.9058356851	0.0479137423
##	7311	7322	7325	7326	7327	7332
##	0.0924853319	0.8513873863	0.6629117540	0.3134090408	0.3119370607	0.4523286347
##	7337	7343	7344	7345	7347	7348
##	0.4678011762	0.8138901632	0.0298942361	0.6052681547	0.0418168566	0.0487252660
##	7351	7355	7357	7363	7366	7367
##	0.1106228809	0.9343116083	0.6331113410	0.4599261419	0.4628384129	0.1345859813
##	7368	7372	7374	7380	7383	7389
##	0.2294276931	0.1280525256	0.9765657938	0.6658840603	0.4735053719	0.0242880834
##	7391	7393	7396	7402	7405	7409
##	0.0998005500	0.3237764933	0.8965277859	0.0387801821	0.9998441945	0.4258805747
##	7415	7418	7425	7428	7433	7435
##	0.9149237997	0.3952719411	0.0002403137	0.1031685121	0.9898557583	0.0547637960
##	7438	7439	7445	7448	7457	7458
##	0.4354302851	0.4015187434	0.9987105964	0.6685971290	0.1184313153	0.2388010404
##	7463	7465	7470	7471	7475	7476
##	0.0968075738	0.6861141918	0.2162114591	0.8330426445	0.2288547026	0.2702654580
##	7483	7484	7487	7489	7491	7493
##	0.3235253563	0.2828863881	0.9452929750	0.4970123188	0.0558496021	0.8286141286
##	7494	7495	7498	7504	7508	7509

##	0.0451569889	0.2587167266	0.2323183369	0.8976062905	0.0222077239	0.7503541504
##	7521	7523	7524	7527	7538	7542
##	0.5377292780	0.8469362104	0.0674959616	0.8682350593	0.0519403742	0.3225453111
##	7546	7547	7553	7560	7570	7571
##	0.8999694613	0.6776658114	0.4870976572	0.0337628137	0.1720970684	0.1254086689
##	7578	7579	7583	7584	7593	7597
##	0.0636413832	0.5113028981	0.0631477279	0.0843888512	0.5739508256	0.8576748617
##	7603	7606	7608	7609	7615	7618
##	0.4099403820	0.6045363422	0.7508515493	0.2637270884	0.1192846811	0.2602935096
##	7620	7628	7630	7631	7637	7638
##	0.3565266034	0.9932838249	0.0775581287	0.1668019845	0.2058025371	0.1287848176
##	7640	7642	7644	7649	7653	7663
##	0.4187711651	0.5553609135	0.9981640822	0.9267866861	0.2064546939	0.2456954756
##	7668	7669	7671	7672	7673	7677
##	0.0883314180	0.3742330307	0.0149279334	0.3957114550	0.6979448057	0.2987606225
##	7682	7684	7686	7687	7691	7694
##	0.7456277072	0.4989326852	0.0008342736	0.6614426637	0.5133048029	0.0638651801
##	7698	7701	7702	7718	7721	7722
##	0.4094875229	0.0657965908	0.0590783622	0.1084583093	0.1495082916	0.4763462175
##	7724	7725	7738	7742	7744	7745
##	0.8978702810	0.2682594897	0.8688226642	0.4563831206	0.5151443783	0.9236938116
##	7747	7756	7760	7767	7768	7771
##	0.8424981042	0.5074891877	0.1827605236	0.8869207598	0.0400784824	0.9964577127
##	7772	7774	7783	7785	7787	7788
##	0.6335399744	0.4690560497	0.0910954635	0.4311157120	0.3577044239	0.6300555274
##	7793	7798	7799	7801	7805	7808
##	0.3043459047	0.1415326779	0.2151965181	0.9960122414	0.5227561525	0.4611415132
##	7813	7815	7821	7824	7829	7832
##	0.7222843383	0.3920625751	0.0888341249	0.1338069273	0.1413685419	0.3744242882
##	7839	7840	7845	7847	7850	7851
##	0.5421373372	0.8706999455	0.3165516219	0.1307840157	0.6383328054	0.0601839524
##	7865	7868	7874	7879	7881	7882
##	0.3853557182	0.2973873778	0.8699408616	0.1874448779	0.8077431044	0.0196950666
##	7884	7886	7889	7890	7893	7902
##	0.1603857389	0.0639259751	0.9973125263	0.2894629597	0.1139301658	0.9807547831
##	7904	7908	7912	7918	7920	7921
##	0.8716877476	0.1719655162	0.9956960456	0.1443797407	0.1275611158	0.0988324645
##	7925	7930	7931	7937	7938	7947
##	0.8977462360	0.0536120114	0.1779234437	0.0070797753	0.1655613212	0.9411788322
##	7948	7952	7960	7962	7970	7972
##	0.6248251078	0.0152342986	0.9924921036	0.4359332985	0.0229615611	0.2733962104
##	7973	7975	7978	7979	7980	7983
##	0.9835287690	0.3907914587	0.0686369665	0.4530486333	0.9908278586	0.9898470761
##	7985	7987	7988	7991	8000	8002
##	0.0464152181	0.9193874110	0.8620981454	0.9977932331	0.1100622464	0.9915733909
##	8015	8019	8021	8025	8033	8036
##	0.9878421703	0.2700104305	0.1343124572	0.6153289732	0.6945166781	0.2177702256
##	8045	8051	8052	8056	8057	8059
##	0.7088399932	0.9988378599	0.2001589371	0.3485668957	0.9061435809	0.4990304442
##	8063	8067	8078	8079	8080	8083
##	0.8013775106	0.4269293624	0.8659621366	0.1064679197	0.0903844865	0.1072217465
##	8087	8091	8096	8099	8104	8108
##	0.0955849775	0.0623618265	0.1413721925	0.1024613990	0.1139896321	0.3008028491
##	8109	8112	8124	8136	8138	8140

```
## 0.9971768410 0.0629836559 0.1672086410 0.0422126014 0.3695951458 0.9438498950
##      8142      8144      8147      8149      8153      8154
## 0.0338398491 0.1023026847 0.4908054877 0.1791693125 0.3574532760 0.9308677143
##      8157      8160      8170      8176      8183      8206
## 0.9876275644 0.4116433363 0.2186167063 0.8168667266 0.9931355052 0.3483357037
##      8207      8209      8211      8216      8230      8241
## 0.4279646482 0.3264669900 0.0851804847 0.2721992022 0.3469934901 0.0690838069
##      8245      8249      8255      8265      8266      8267
## 0.1369898711 0.2762762765 0.1515746773 0.0001645700 0.1251792783 0.9458550007
##      8273      8284      8285      8289      8294      8297
## 0.9594486889 0.4806863704 0.7816911580 0.3610272184 0.2071575643 0.0285060588
##      8303      8305      8319      8320      8322      8325
## 0.0679959682 0.6294381900 0.9420492102 0.4661849021 0.9099086629 0.0390349603
##      8330      8335      8339      8340      8341      8349
## 0.6503877532 0.5615915068 0.0740465081 0.6844706395 0.7804705436 0.1737888669
##      8357      8358      8366      8371      8372      8373
## 0.7203391739 0.0643430160 0.2075303929 0.0363743900 0.3296286250 0.1601625074
##      8374      8386      8389      8396      8399      8412
## 0.6786673083 0.0565757117 0.1994286309 0.0440869935 0.1948840740 0.2084650583
##      8413      8425      8429      8434      8438      8440
## 0.1497561978 0.1381987496 0.7798062316 0.9925426483 0.7790550780 0.9303494516
##      8446      8451      8453      8454      8462      8464
## 0.0412895406 0.6652936830 0.1301928336 0.9999434721 0.2246600373 0.5193171953
##      8469      8472      8473      8479      8481      8487
## 0.2735949964 0.3979867737 0.4832977107 0.5170726567 0.4896236159 0.0811328246
##      8491      8494      8499      8501      8504      8514
## 0.0389597954 0.5144790599 0.4742996031 0.0254169161 0.7551348045 0.0820815406
##      8521      8526
## 0.6908998166 0.3197976914
```

```
y_pred = ifelse(prob_pred > 0.5, 1, 0)
y_pred
```

```
##      2      5      6      7     11     15     16     18     20     21     31     36     37     41     42     43
##      1      1      1      0      0      0      0      0      1      0      1      1      1      0      0      1
##     56     59     63     64     67     69     83     86     89    101    110    115    124    125    128    140
##      1      0      1      0      0      0      0      0      0      1      0      0      1      0      0      0
##    143    148    149    151    154    155    158    160    163    164    165    167    169    177    179    182
##      0      0      1      0      1      0      1      0      1      0      0      0      1      0      1      0
##    183    187    194    197    201    202    205    211    217    218    220    225    226    229    236    245
##      0      0      0      0      0      0      0      1      1      1      0      0      0      0      0      0
##    247    252    254    256    261    264    267    270    274    279    283    290    291    293    298    301
##      0      0      0      0      0      0      0      1      0      1      0      0      0      0      0      0
##    303    314    321    322    323    324    326    334    339    341    346    349    350    351    353    365
##      1      1      1      0      0      1      0      0      0      0      0      0      0      0      0      0
##    366    367    371    374    383    385    389    397    399    404    405    406    407    413    414    420
##      0      0      0      0      0      0      1      0      1      0      1      1      0      0      1      1
##    421    422    438    439    445    446    449    450    456    464    468    474    476    481    485    499
##      1      0      1      1      0      0      1      0      1      1      1      1      0      1      1      0
##    500    506    509    512    513    520    524    528    531    532    535    537    540    551    552    554
##      1      1      1      0      1      0      0      0      1      0      0      0      1      1      0      1
##    555    559    560    570    573    574    587    589    598    599    603    604    605    607    608    616
##      0      0      0      0      1      1      0      0      1      1      0      0      0      1      1      1
##    622    626    633    634    637    638    649    654    655    656    657    662    666    667    668    669
##      1      1      0      0      1      0      0      0      0      0      0      0      1      0      1      1
```

##	673	674	675	676	682	683	685	689	692	705	711	717	719	721	725	726
##	0	0	0	1	1	0	1	0	0	0	0	0	1	1	0	1
##	727	729	731	734	737	741	745	751	756	762	766	768	774	775	778	780
##	0	1	0	0	0	0	1	0	1	0	0	0	1	1	1	1
##	784	788	794	796	797	802	803	808	816	818	821	822	827	829	832	838
##	0	0	0	0	1	0	0	0	1	1	0	0	0	0	1	0
##	842	848	849	860	861	865	870	876	878	880	881	887	888	892	893	900
##	1	0	0	0	0	1	1	0	0	0	0	0	1	1	0	0
##	901	907	913	915	919	922	923	925	927	931	936	944	951	953	954	958
##	1	0	0	0	1	0	0	0	0	0	1	0	0	0	0	1
##	959	960	962	964	967	971	982	998	999	1003	1008	1013	1026	1028	1034	1042
##	0	1	0	1	1	0	0	0	0	1	0	1	0	0	0	0
##	1047	1049	1050	1056	1062	1069	1070	1081	1085	1089	1093	1096	1111	1119	1129	1131
##	0	1	1	1	0	0	0	0	0	0	1	1	1	0	1	0
##	1133	1134	1138	1140	1142	1146	1147	1148	1153	1156	1162	1167	1170	1175	1176	1182
##	0	0	0	1	0	0	0	0	0	0	1	0	0	0	0	0
##	1185	1193	1198	1202	1213	1214	1225	1232	1240	1241	1244	1249	1252	1257	1262	1266
##	0	0	0	0	1	1	0	1	0	0	0	0	1	1	0	0
##	1269	1270	1278	1284	1287	1290	1294	1303	1305	1313	1318	1319	1329	1330	1335	1338
##	0	1	0	0	0	1	0	0	1	1	1	0	1	0	0	1
##	1339	1341	1344	1347	1349	1356	1357	1364	1367	1368	1373	1374	1375	1379	1383	1385
##	0	1	0	0	1	0	0	0	1	1	0	0	0	1	1	0
##	1389	1396	1404	1406	1407	1408	1410	1412	1419	1421	1423	1427	1428	1429	1432	1436
##	1	1	0	0	0	1	0	1	0	1	0	1	0	1	0	1
##	1443	1446	1451	1453	1457	1463	1466	1473	1475	1476	1477	1478	1479	1481	1483	1484
##	0	0	1	1	0	0	1	0	1	1	0	0	0	1	1	1
##	1485	1487	1492	1496	1499	1501	1507	1508	1512	1516	1520	1523	1527	1529	1533	1546
##	0	0	1	0	0	1	0	0	1	1	1	1	1	0	1	0
##	1547	1548	1552	1553	1558	1561	1565	1569	1571	1577	1580	1581	1599	1619	1621	1623
##	0	1	0	0	0	1	0	1	0	0	1	1	0	0	1	0
##	1634	1638	1640	1641	1645	1652	1665	1673	1677	1683	1689	1694	1718	1719	1723	1727
##	1	0	0	0	0	0	1	0	1	0	0	0	1	0	0	0
##	1729	1730	1731	1735	1736	1737	1740	1743	1746	1751	1752	1754	1760	1761	1763	1767
##	1	0	0	1	0	1	0	0	1	1	0	0	0	1	0	1
##	1769	1782	1783	1792	1796	1803	1806	1807	1808	1816	1821	1823	1824	1827	1830	1837
##	0	0	0	0	0	0	1	1	1	1	0	0	0	0	0	0
##	1839	1840	1843	1846	1849	1855	1856	1858	1864	1866	1870	1872	1876	1878	1882	1891
##	1	0	0	0	1	0	0	0	0	0	0	1	1	0	1	0
##	1893	1900	1904	1925	1926	1928	1929	1933	1939	1943	1956	1964	1966	1972	1975	1976
##	1	1	1	0	1	1	1	1	1	1	1	0	0	1	0	0
##	1981	1991	1999	2003	2004	2007	2008	2015	2017	2018	2020	2027	2028	2029	2035	2042
##	0	0	0	0	0	1	0	0	0	0	1	0	1	0	1	0
##	2046	2049	2050	2052	2053	2056	2073	2074	2076	2081	2083	2084	2088	2093	2098	2102
##	1	0	0	0	0	0	0	0	1	0	1	0	0	0	0	0
##	2107	2113	2115	2117	2120	2134	2135	2139	2140	2145	2150	2152	2154	2155	2162	2167
##	1	1	0	0	0	0	1	0	1	1	0	0	0	1	0	1
##	2168	2174	2175	2188	2198	2199	2203	2216	2217	2218	2222	2226	2230	2235	2241	2247
##	0	1	1	0	0	0	0	0	1	0	0	0	1	0	1	0
##	2253	2254	2255	2256	2259	2264	2270	2272	2275	2277	2278	2282	2283	2289	2309	2315
##	1	1	0	0	1	1	0	1	0	0	0	0	0	0	1	0
##	2316	2322	2327	2329	2330	2332	2338	2345	2349	2353	2356	2357	2365	2373	2376	2377
##	1	1	0	0	0	0	0	1	0	0	1	0	0	0	0	0
##	2385	2389	2391	2392	2394	2400	2403	2406	2408	2412	2421	2426	2430	2431	2432	2436
##	1	0	1	1	0	0	0	0	1	0	1	0	0	0	1	1

##	2439	2447	2457	2460	2462	2464	2465	2466	2469	2474	2475	2479	2486	2488	2491	2504
##	0	0	1	1	0	0	0	0	0	0	0	1	1	0	0	0
##	2505	2510	2514	2515	2516	2520	2521	2523	2526	2531	2534	2535	2537	2538	2541	2542
##	0	1	0	1	0	0	1	0	0	0	1	1	0	1	0	0
##	2546	2554	2563	2574	2579	2580	2581	2582	2586	2591	2594	2595	2598	2601	2610	2615
##	0	0	1	1	1	0	0	0	0	0	1	1	0	0	0	0
##	2619	2626	2627	2631	2632	2639	2640	2644	2645	2647	2649	2656	2657	2658	2659	2661
##	0	0	0	0	1	0	0	0	1	0	1	0	1	0	0	0
##	2662	2669	2672	2675	2679	2688	2693	2695	2696	2699	2703	2704	2712	2722	2736	2744
##	0	0	0	1	0	0	0	0	0	1	1	1	0	0	0	1
##	2746	2747	2751	2753	2754	2755	2758	2763	2767	2768	2769	2770	2771	2780	2784	2785
##	1	1	0	0	0	1	0	0	0	1	0	0	0	1	0	1
##	2787	2793	2798	2803	2804	2807	2810	2812	2820	2825	2832	2834	2839	2843	2848	2850
##	0	0	1	1	1	0	0	0	0	0	0	0	0	0	0	1
##	2856	2866	2872	2873	2878	2879	2883	2892	2899	2900	2902	2905	2906	2915	2916	2923
##	0	0	0	0	1	1	0	0	1	0	0	1	0	1	1	0
##	2927	2934	2936	2940	2943	2945	2949	2951	2957	2959	2962	2963	2965	2967	2981	2984
##	1	1	0	1	1	0	0	1	0	0	1	0	0	0	1	0
##	2985	2989	3004	3011	3017	3028	3037	3042	3044	3050	3058	3059	3061	3064	3067	3068
##	0	0	1	0	1	1	1	0	1	0	0	0	0	0	0	1
##	3071	3074	3075	3076	3082	3089	3093	3095	3098	3101	3102	3105	3106	3111	3119	3127
##	1	0	0	0	1	0	0	0	0	0	1	0	1	0	1	0
##	3134	3136	3137	3147	3151	3160	3172	3173	3180	3181	3185	3186	3189	3193	3196	3198
##	0	0	1	0	1	0	1	0	0	1	0	0	0	0	0	1
##	3204	3206	3210	3212	3217	3227	3231	3233	3238	3240	3243	3245	3246	3249	3254	3257
##	1	0	0	0	0	0	0	0	1	1	0	0	0	0	0	0
##	3259	3265	3272	3282	3285	3289	3294	3305	3310	3323	3325	3328	3329	3330	3339	3341
##	1	0	0	0	1	0	1	0	0	0	0	0	0	1	1	0
##	3344	3348	3350	3354	3356	3357	3359	3363	3364	3374	3384	3387	3395	3399	3402	3405
##	0	0	1	0	0	1	0	0	1	1	0	0	0	0	0	1
##	3408	3409	3416	3417	3419	3427	3430	3431	3432	3434	3435	3438	3441	3450	3461	3462
##	0	0	0	1	0	1	0	0	0	0	0	1	1	0	1	1
##	3464	3466	3467	3477	3478	3483	3490	3493	3499	3503	3506	3510	3514	3523	3524	3533
##	1	1	1	1	0	1	0	0	0	1	0	0	0	0	0	1
##	3534	3541	3542	3544	3545	3548	3549	3551	3558	3559	3563	3579	3580	3583	3585	3587
##	0	1	0	0	1	0	1	1	1	0	0	1	1	0	1	1
##	3590	3591	3592	3596	3597	3598	3602	3603	3608	3610	3613	3620	3621	3624	3626	3630
##	1	0	1	1	0	0	1	1	1	1	1	0	0	0	0	1
##	3631	3641	3646	3647	3650	3654	3659	3660	3664	3676	3689	3692	3695	3700	3706	3707
##	0	0	0	0	0	0	1	1	0	0	1	0	1	0	1	1
##	3711	3714	3715	3724	3726	3727	3741	3742	3749	3755	3756	3758	3761	3763	3766	3775
##	1	0	1	0	1	1	0	0	0	1	0	0	0	0	1	0
##	3778	3779	3781	3782	3789	3795	3801	3808	3819	3823	3824	3833	3836	3839	3841	3846
##	1	1	1	1	0	1	0	0	0	0	1	0	0	1	0	1
##	3847	3851	3872	3877	3878	3881	3882	3883	3893	3895	3896	3898	3899	3900	3906	3914
##	0	0	1	1	1	0	1	0	0	0	1	0	0	1	0	0
##	3917	3924	3927	3928	3937	3938	3941	3949	3951	3952	3958	3960	3962	3963	3965	3967
##	0	1	0	1	1	0	0	0	1	0	0	0	0	1	0	0
##	3969	3970	3972	3973	3974	3981	3982	3993	3996	3997	4002	4004	4006	4007	4015	4017
##	0	0	1	1	0	0	0	0	0	0	1	0	1	0	0	0
##	4018	4021	4033	4034	4047	4048	4056	4060	4065	4066	4069	4075	4076	4077	4078	4083
##	1	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0
##	4088	4090	4092	4095	4097	4100	4108	4109	4110	4112	4113	4122	4127	4131	4134	4135
##	0	1	0	0	0	0	0	0	1	1	0	0	1	0	0	0

##	4136	4138	4139	4140	4150	4151	4152	4157	4158	4162	4166	4167	4169	4171	4175	4178
##	1	0	0	0	0	0	0	0	0	0	0	0	1	0	1	1
##	4184	4185	4190	4191	4197	4204	4205	4208	4209	4210	4212	4215	4217	4218	4219	4225
##	0	0	0	1	0	0	0	0	0	0	0	0	0	1	1	1
##	4236	4238	4241	4243	4245	4247	4248	4249	4256	4259	4266	4269	4271	4280	4282	4283
##	1	1	1	0	1	1	0	0	0	1	0	0	1	1	1	0
##	4287	4290	4291	4293	4317	4326	4329	4333	4334	4337	4340	4345	4353	4359	4364	4366
##	0	1	0	1	0	0	0	0	0	0	0	0	1	0	0	1
##	4371	4374	4377	4378	4379	4381	4388	4390	4391	4396	4403	4407	4409	4413	4415	4417
##	0	1	0	0	1	0	1	0	0	0	0	0	1	1	0	0
##	4419	4423	4424	4427	4428	4431	4433	4438	4439	4440	4444	4445	4448	4456	4461	4464
##	1	1	1	0	0	0	0	0	0	0	1	1	0	0	1	0
##	4469	4479	4481	4482	4484	4485	4486	4489	4492	4498	4502	4505	4507	4509	4510	4512
##	0	0	0	0	1	0	0	0	0	0	0	0	0	0	1	0
##	4517	4520	4523	4524	4526	4533	4534	4540	4542	4548	4552	4560	4572	4578	4582	4585
##	1	0	1	0	0	0	1	1	1	0	0	1	1	1	1	0
##	4596	4625	4630	4633	4642	4654	4656	4670	4674	4678	4688	4693	4698	4703	4715	4716
##	1	0	1	0	1	0	0	0	1	1	1	0	1	1	0	1
##	4722	4723	4731	4739	4742	4746	4748	4753	4757	4759	4763	4771	4775	4778	4780	4783
##	0	1	0	0	1	1	1	1	0	0	0	1	0	0	0	0
##	4784	4786	4787	4788	4801	4806	4807	4811	4817	4826	4831	4840	4845	4854	4858	4863
##	1	0	0	0	1	1	0	1	0	0	0	1	1	0	1	1
##	4868	4870	4871	4877	4879	4889	4890	4891	4899	4902	4905	4906	4908	4913	4917	4919
##	0	0	1	1	0	0	0	0	1	0	0	0	0	0	0	0
##	4924	4925	4926	4927	4935	4936	4937	4938	4939	4940	4946	4947	4950	4959	4965	4966
##	1	0	0	0	0	1	0	0	0	1	0	0	0	0	0	1
##	4968	4970	4972	4978	4980	4984	4987	4988	4989	4991	5000	5003	5006	5009	5011	5014
##	0	0	0	1	0	1	0	0	1	0	0	0	0	0	1	0
##	5016	5017	5020	5021	5029	5039	5041	5051	5056	5059	5064	5070	5074	5075	5087	5089
##	1	0	1	0	1	0	0	0	1	1	0	1	1	1	0	0
##	5093	5095	5100	5101	5116	5123	5124	5127	5131	5133	5134	5137	5138	5140	5144	5153
##	0	0	1	1	1	1	0	0	1	1	0	1	1	1	0	0
##	5154	5163	5166	5172	5186	5188	5189	5193	5194	5197	5204	5206	5211	5214	5215	5216
##	1	0	0	0	1	0	0	1	0	0	0	1	1	0	0	0
##	5218	5226	5227	5230	5236	5244	5245	5248	5250	5251	5254	5261	5262	5264	5265	5268
##	1	0	1	1	0	1	0	0	0	0	0	0	1	1	1	0
##	5269	5273	5275	5276	5280	5283	5284	5294	5295	5296	5300	5303	5305	5310	5330	5335
##	0	0	0	0	1	1	1	0	0	1	0	0	1	1	0	0
##	5352	5353	5355	5357	5358	5359	5360	5363	5365	5366	5369	5371	5376	5378	5382	5385
##	0	0	1	0	1	1	0	0	1	0	0	0	0	1	0	0
##	5389	5393	5400	5402	5405	5406	5407	5410	5411	5417	5418	5419	5431	5432	5433	5442
##	1	0	1	0	1	0	1	0	0	0	1	1	0	1	1	1
##	5449	5452	5454	5455	5462	5466	5472	5473	5474	5483	5491	5495	5503	5508	5510	5511
##	0	0	1	0	1	0	1	1	1	0	0	1	1	0	1	1
##	5513	5515	5520	5524	5529	5539	5545	5550	5551	5554	5566	5584	5589	5599	5602	5604
##	0	0	1	1	1	0	0	0	0	1	0	0	0	0	1	0
##	5607	5608	5614	5616	5621	5622	5628	5638	5642	5651	5654	5662	5666	5675	5676	5691
##	1	1	0	0	0	1	0	0	1	0	0	0	0	0	1	0
##	5693	5704	5707	5709	5710	5715	5718	5720	5723	5726	5729	5732	5736	5739	5740	5745
##	1	1	0	1	0	0	0	1	0	1	0	1	0	0	0	1
##	5746	5751	5752	5757	5758	5759	5765	5770	5772	5777	5779	5780	5782	5797	5806	5810
##	1	0	1	0	1	0	0	0	1	0	1	0	0	1	0	0
##	5811	5814	5820	5828	5835	5849	5857	5862	5865	5867	5868	5871	5875	5878	5881	5882
##	1	0	0	0	1	1	1	0	0	1	0	1	0	0	0	0

##	5887	5893	5901	5904	5909	5911	5922	5923	5924	5925	5930	5931	5932	5934	5937	5941
##	0	1	0	0	0	0	0	0	1	0	0	1	1	0	1	0
##	5946	5949	5951	5953	5963	5973	5977	5978	5980	5984	5993	5995	5998	6000	6004	6008
##	1	1	1	0	0	0	0	1	1	1	0	1	1	0	0	0
##	6015	6016	6018	6030	6035	6044	6047	6049	6050	6052	6056	6057	6059	6063	6065	6066
##	1	1	1	0	0	1	0	0	1	1	0	0	0	0	0	0
##	6068	6069	6076	6096	6097	6099	6100	6103	6106	6109	6115	6116	6117	6119	6121	6125
##	0	0	0	1	0	1	1	1	1	1	0	0	0	0	0	1
##	6129	6142	6143	6146	6148	6158	6159	6167	6168	6169	6177	6185	6190	6193	6196	6199
##	1	0	0	0	0	0	1	0	0	1	0	1	1	1	0	1
##	6200	6206	6211	6219	6221	6222	6223	6224	6228	6235	6237	6242	6246	6250	6258	6264
##	1	1	0	1	1	0	1	0	0	0	1	0	0	0	0	0
##	6265	6268	6272	6275	6284	6285	6290	6293	6294	6296	6306	6317	6324	6325	6328	6334
##	0	0	1	0	0	0	0	0	0	0	1	0	1	1	0	0
##	6337	6340	6344	6348	6350	6352	6353	6354	6358	6360	6361	6362	6363	6379	6380	6391
##	0	0	0	0	0	0	0	1	0	1	0	1	1	0	0	0
##	6393	6397	6400	6411	6415	6418	6419	6422	6423	6430	6437	6440	6444	6451	6460	6462
##	0	1	0	1	0	0	0	0	1	0	1	0	0	1	0	0
##	6463	6465	6466	6467	6469	6471	6477	6478	6481	6483	6485	6486	6487	6488	6489	6490
##	0	1	0	1	1	0	1	0	1	0	0	0	1	0	0	0
##	6491	6492	6497	6499	6502	6508	6515	6521	6522	6527	6532	6534	6535	6538	6549	6551
##	0	0	0	0	1	0	1	1	0	1	1	0	0	1	1	0
##	6553	6555	6559	6561	6563	6565	6566	6569	6575	6580	6582	6586	6614	6615	6617	6618
##	1	0	0	1	0	1	0	0	0	1	1	0	0	1	0	0
##	6620	6632	6646	6654	6658	6661	6663	6664	6665	6673	6677	6681	6682	6686	6687	6690
##	0	0	0	0	0	0	0	1	0	0	0	1	1	1	0	1
##	6693	6694	6695	6697	6698	6702	6705	6717	6724	6727	6728	6730	6732	6733	6738	6740
##	1	0	1	1	0	0	1	1	1	0	0	1	1	0	1	0
##	6743	6751	6756	6757	6760	6763	6767	6770	6771	6777	6782	6783	6784	6788	6790	6793
##	1	0	0	0	1	1	0	0	0	1	0	0	0	0	0	1
##	6797	6798	6803	6807	6813	6815	6825	6832	6837	6848	6850	6852	6855	6859	6864	6868
##	0	0	1	0	1	0	0	1	1	0	0	0	0	0	1	0
##	6871	6873	6885	6887	6891	6895	6900	6903	6909	6920	6922	6926	6930	6932	6941	6952
##	1	1	0	0	0	1	0	0	1	0	0	0	0	0	0	0
##	6953	6958	6961	6964	6974	6976	6977	6978	6979	6981	6986	6993	7002	7005	7015	7020
##	0	1	0	0	0	0	0	0	0	1	0	1	0	1	0	1
##	7022	7026	7028	7039	7043	7047	7058	7073	7074	7081	7083	7084	7085	7086	7087	7089
##	0	0	0	0	0	0	1	0	1	0	0	0	0	1	0	1
##	7094	7097	7098	7104	7110	7115	7118	7124	7129	7133	7138	7146	7155	7160	7164	7165
##	0	0	0	0	0	1	1	0	1	1	0	0	0	0	1	0
##	7169	7172	7173	7180	7187	7194	7196	7200	7202	7203	7204	7211	7215	7217	7220	7226
##	1	0	0	1	0	0	1	0	1	0	0	0	1	0	0	0
##	7237	7240	7241	7244	7248	7249	7254	7263	7264	7265	7267	7268	7271	7273	7274	7277
##	1	1	0	0	0	1	1	1	0	0	0	0	1	0	0	0
##	7278	7283	7289	7295	7298	7300	7311	7322	7325	7326	7327	7332	7337	7343	7344	7345
##	1	0	0	0	1	0	0	1	1	0	0	0	0	1	0	1
##	7347	7348	7351	7355	7357	7363	7366	7367	7368	7372	7374	7380	7383	7389	7391	7393
##	0	0	0	1	1	0	0	0	0	0	1	1	0	0	0	0
##	7396	7402	7405	7409	7415	7418	7425	7428	7433	7435	7438	7439	7445	7448	7457	7458
##	1	0	1	0	1	0	0	0	1	0	0	0	1	1	0	0
##	7463	7465	7470	7471	7475	7476	7483	7484	7487	7489	7491	7493	7494	7495	7498	7504
##	0	1	0	1	0	0	0	0	1	0	0	1	0	0	0	1
##	7508	7509	7521	7523	7524	7527	7538	7542	7546	7547	7553	7560	7570	7571	7578	7579
##	0	1	1	1	0	1	0	0	1	1	0	0	0	0	0	1

```
## 7583 7584 7593 7597 7603 7606 7608 7609 7615 7618 7620 7628 7630 7631 7637 7638
##      0      0      1      1      0      1      1      0      0      0      0      1      0      0      0      0
## 7640 7642 7644 7649 7653 7663 7668 7669 7671 7672 7673 7677 7682 7684 7686 7687
##      0      1      1      1      0      0      0      0      0      0      1      0      1      0      0      1
## 7691 7694 7698 7701 7702 7718 7721 7722 7724 7725 7738 7742 7744 7745 7747 7756
##      1      0      0      0      0      0      0      0      1      0      1      0      1      1      1      1
## 7760 7767 7768 7771 7772 7774 7783 7785 7787 7788 7793 7798 7799 7801 7805 7808
##      0      1      0      1      1      0      0      0      0      1      0      0      0      1      1      0
## 7813 7815 7821 7824 7829 7832 7839 7840 7845 7847 7850 7851 7865 7868 7874 7879
##      1      0      0      0      0      0      1      1      0      0      1      0      0      0      1      0
## 7881 7882 7884 7886 7889 7890 7893 7902 7904 7908 7912 7918 7920 7921 7925 7930
##      1      0      0      0      1      0      0      1      1      0      1      0      0      0      1      0
## 7931 7937 7938 7947 7948 7952 7960 7962 7970 7972 7973 7975 7978 7979 7980 7983
##      0      0      0      1      1      0      1      0      0      0      1      0      0      0      1      1
## 7985 7987 7988 7991 8000 8002 8015 8019 8021 8025 8033 8036 8045 8051 8052 8056
##      0      1      1      1      0      1      1      0      0      1      1      0      1      1      0      0
## 8057 8059 8063 8067 8078 8079 8080 8083 8087 8091 8096 8099 8104 8108 8109 8112
##      1      0      1      0      1      0      0      0      0      0      0      0      0      0      1      0
## 8124 8136 8138 8140 8142 8144 8147 8149 8153 8154 8157 8160 8170 8176 8183 8206
##      0      0      0      1      0      0      0      0      0      1      1      0      0      1      1      0
## 8207 8209 8211 8216 8230 8241 8245 8249 8255 8265 8266 8267 8273 8284 8285 8289
##      0      0      0      0      0      0      0      0      0      0      0      1      1      0      1      0
## 8294 8297 8303 8305 8319 8320 8322 8325 8330 8335 8339 8340 8341 8349 8357 8358
##      0      0      0      1      1      0      1      0      1      1      0      1      1      0      1      0
## 8366 8371 8372 8373 8374 8386 8389 8396 8399 8412 8413 8425 8429 8434 8438 8440
##      0      0      0      0      1      0      0      0      0      0      0      0      1      1      1      1
## 8446 8451 8453 8454 8462 8464 8469 8472 8473 8479 8481 8487 8491 8494 8499 8501
##      0      1      0      1      0      1      0      0      0      1      0      0      0      1      0      0
## 8504 8514 8521 8526
##      1      0      1      0
```

```
#making the confusion matrix
```

```
cm = table(test_set[,17], y_pred)
cm
```

```
##      y_pred
##      0      1
## 0 1110 130
## 1 257 635
```

```
print("=====Logistic Regression=====")
```

```
## [1] "=====Logistic Regression====="
```

```
library(ggplot2)
library(lattice)
library(caret)
confusionMatrix(cm)
```

```
## Confusion Matrix and Statistics
##
##      y_pred
##      0      1
## 0 1110 130
## 1 257 635
##
```



```
##              Accuracy : 0.8185
##              95% CI : (0.8015, 0.8346)
##      No Information Rate : 0.6412
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.6194
##
##      McNemar's Test P-Value : 1.504e-10
##
##              Sensitivity : 0.8120
##              Specificity : 0.8301
##      Pos Pred Value : 0.8952
##      Neg Pred Value : 0.7119
##              Prevalence : 0.6412
##      Detection Rate : 0.5206
##      Detection Prevalence : 0.5816
##      Balanced Accuracy : 0.8210
##
##      'Positive' Class : 0
##
```

```
#####
#Naive Bayes
rm(list=ls())
# Importing the dataset
dataset = read.csv('train.csv')

dataset$days_elapsed_old[dataset$days_elapsed_old<1] <- 0
dataset[ dataset == "na" ] <- NA

#Factor like columns
dataset$job=as.integer(as.factor(dataset$job))
dataset$marital=as.integer(as.factor(dataset$marital))
dataset$education=as.integer(as.factor(dataset$education))
dataset$device=as.integer(as.factor(dataset$device))
dataset$outcome_old=as.integer(as.factor(dataset$outcome_old))
dataset[is.na(dataset)] <- 0

# Encoding the target feature as factor
dataset$y = factor(dataset$y, levels = c(0, 1)) #labels /levels -both are same

# Splitting the dataset into the Training set and Test set
# install.packages('caTools')
library(caTools)
set.seed(123)
split = sample.split(dataset$y, SplitRatio = 0.75)
training_set = subset(dataset, split == TRUE)
test_set = subset(dataset, split == FALSE)

# Feature Scaling
training_set[-17] = scale(training_set[-17])
test_set[-17] = scale(test_set[-17])

# Fitting Naive Bayes to the Training set
```

```

library(e1071)
classifier = naiveBayes(x = training_set[-17],
                        y = training_set$y)

# Predicting the Test set results
y_pred = predict(classifier, newdata = test_set[-17])

# Making the Confusion Matrix
cm = table(test_set[, 17], y_pred)

print("=====Naive Bayes=====")

## [1] "=====Naive Bayes=====
library(ggplot2)
library(lattice)
library(caret)
confusionMatrix(cm)

## Confusion Matrix and Statistics
##
##      y_pred
##      0      1
## 0 1020  220
## 1   325  567
##
##              Accuracy : 0.7444
##              95% CI : (0.7253, 0.7628)
##      No Information Rate : 0.6309
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.4659
##
##  Mcnemar's Test P-Value : 8.394e-06
##
##              Sensitivity : 0.7584
##              Specificity : 0.7205
##              Pos Pred Value : 0.8226
##              Neg Pred Value : 0.6357
##              Prevalence : 0.6309
##              Detection Rate : 0.4784
##      Detection Prevalence : 0.5816
##              Balanced Accuracy : 0.7394
##
##              'Positive' Class : 0
##
#=====
#Decision Tree Classification

rm(list=ls())

# Importing the dataset
dataset = read.csv('train.csv')

```

```

dataset$days_elapsed_old[dataset$days_elapsed_old<1] <- 0
dataset[ dataset == "na" ] <- NA

#Factor like columns
dataset$job=as.integer(as.factor(dataset$job))
dataset$marital=as.integer(as.factor(dataset$marital))
dataset$education=as.integer(as.factor(dataset$education))
dataset$device=as.integer(as.factor(dataset$device))
dataset$outcome_old=as.integer(as.factor(dataset$outcome_old))
dataset[is.na(dataset)] <- 0

# Encoding the target feature as factor
dataset$y = factor(dataset$y, levels = c(0, 1))

# Splitting the dataset into the Training set and Test set
# install.packages('caTools')
library(caTools)
set.seed(123)
split = sample.split(dataset$y, SplitRatio = 0.75)
training_set = subset(dataset, split == TRUE)
test_set = subset(dataset, split == FALSE)

# Feature Scaling #no need to scale,but to visualise in high resolution if we scale, the results will b
training_set[-17] = scale(training_set[-17])
test_set[-17] = scale(test_set[-17])

# Fitting Decision TreeClassification to the Training set
library(rpart)
classifier = rpart(formula = y ~ .,
                    data = training_set)

# Predicting the Test set results
y_pred = predict(classifier, newdata = test_set[-17], type = 'class')

# Making the Confusion Matrix
cm = table(test_set[, 17], y_pred)

print("=====Decision Trees=====")

## [1] "=====Decision Trees=====

library(ggplot2)
library(lattice)
library(caret)
confusionMatrix(cm)

## Confusion Matrix and Statistics
##
##      y_pred
##      0      1
## 0 1064  176
## 1   262  630
##
##               Accuracy : 0.7946

```

```
##          95% CI : (0.7768, 0.8115)
##      No Information Rate : 0.622
##      P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.5721
##
##      McNemar's Test P-Value : 4.877e-05
##
##          Sensitivity : 0.8024
##          Specificity : 0.7816
##      Pos Pred Value : 0.8581
##      Neg Pred Value : 0.7063
##          Prevalence : 0.6220
##      Detection Rate : 0.4991
##      Detection Prevalence : 0.5816
##      Balanced Accuracy : 0.7920
##
##      'Positive' Class : 0
##
```

```
#=====
# k-nearest neighbors (K-NN)

rm(list=ls())

# Importing the dataset
dataset = read.csv('train.csv')

dataset$days_elapsed_old[dataset$days_elapsed_old<1] <- 0
dataset[ dataset == "na" ] <- NA

#Factor like columns
dataset$job=as.integer(as.factor(dataset$job))
dataset$marital=as.integer(as.factor(dataset$marital))
dataset$education=as.integer(as.factor(dataset$education))
dataset$device=as.integer(as.factor(dataset$device))
dataset$outcome_old=as.integer(as.factor(dataset$outcome_old))
dataset[is.na(dataset)] <- 0

# Encoding the target feature as factor
# Encoding the target feature as factor #(the values are considered as numeric values i.e 1 > 0 but we
#Instead we want them to consider as factors i.e 1 and 0 as two different categories.)
dataset$y = factor(dataset$y, levels = c(0, 1))

# Splitting the dataset into the Training set and Test set
# install.packages('caTools')
library(caTools)
set.seed(123)
split = sample.split(dataset$y, SplitRatio = 0.75)
training_set = subset(dataset, split == TRUE)
test_set = subset(dataset, split == FALSE)

# Feature Scaling
training_set[-17] = scale(training_set[-17])
```

```

test_set[-17] = scale(test_set[-17])

# Fitting K-NN to the Training set and predicting the test set results
#install.packages('class')
library(class)
y_pred = knn(train = training_set[, -17],
              test = test_set[, -17],
              cl = training_set[, 17], k = 20)
y_pred

##      [1] 1 1 1 0 0 0 0 0 1 0 1 1 1 0 0 1 0 0 1 0 0 0 0 0 0 1 0 0 1 0 0 0 0 0 0 1
##     [38] 0 1 0 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 0 1 0 0 0 1 0 1 0 0 1 0 1
##     [75] 0 0 0 0 0 0 1 1 1 1 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0 1
##    [112] 1 0 0 1 0 0 0 1 0 0 0 1 1 0 0 0 0 0 0 1 1 0 0 0 0 0 1 0 0 0 1 0 0 1 0 0 0 0
##    [149] 1 1 0 0 1 0 0 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 0 1 0 0 1 0 0 0 1 1 0 1 0 0
##    [186] 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 0 0 0 1 1 1 1 0 0 0 0 1 0 0 0 1 1 0 0 0 0
##    [223] 1 0 0 0 0 1 0 0 1 0 0 0 0 0 1 1 0 0 1 0 0 0 1 0 0 0 0 0 0 1 0 0 0 1 0 1 0
##    [260] 1 1 0 0 0 0 1 0 1 1 0 1 0 0 1 1 1 0 0 0 0 0 0 1 1 1 0 1 1 1 0 1 0 0 0 0 1
##    [297] 0 0 1 1 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 0 0 1 1 0 0 0 0 0 0 1 1 1 0 0 1 1 0 1
##    [334] 0 0 1 0 1 1 0 1 0 0 0 1 0 0 0 0 1 0 0 1 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 1 0
##    [371] 1 1 0 0 1 0 1 1 0 0 0 1 1 0 0 0 1 0 0 0 0 0 1 1 1 1 0 1 0 0 0 0 0 0 0 1 0
##    [408] 0 0 0 1 1 1 0 1 0 1 0 0 1 0 0 0 1 0 0 0 0 1 0 0 0 0 0 0 1 0 0 1 0 1 0 0 1
##    [445] 0 1 0 1 0 0 0 1 0 0 1 1 0 0 0 0 1 0 0 0 0 0 0 1 0 0 0 0 0 0 1 1 0 1 0 1
##    [482] 1 1 0 0 1 0 1 1 0 1 0 0 0 0 0 1 1 0 1 0 1 0 0 0 0 0 0 1 0 1 0 1 0 0 0 0
##    [519] 0 0 1 0 1 0 0 0 0 0 1 0 0 0 0 0 1 1 1 1 0 0 1 1 0 1 0 1 1 0 0 1 0 0 0 0
##    [556] 0 1 0 0 0 1 1 0 1 1 0 0 1 0 0 0 0 0 1 0 1 0 0 0 0 0 1 1 0 0 1 0 0 0 1 0
##    [593] 1 0 1 0 0 0 0 0 1 0 1 0 0 1 0 1 0 0 1 1 1 0 0 0 0 0 0 1 1 0 1 0 0 1 0 1 0
##    [630] 1 0 0 0 0 1 1 0 0 0 1 0 0 1 1 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 0 1 0 0 0 1 0
##    [667] 1 0 1 1 0 0 0 0 0 1 0 0 0 1 0 1 1 0 0 0 0 1 1 0 0 0 0 1 0 0 0 1 0 0 0 0
##    [704] 1 0 0 1 1 0 0 0 0 0 0 0 0 1 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 1 0 1 0 1 1 0
##    [741] 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 1 0 0 1 1 1 0 1 0 0 0 0 0 0 1 1 1 0 0 1
##    [778] 0 1 1 0 0 1 0 0 0 1 0 0 0 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 0 0 0 1 1 0 0 0
##    [815] 0 0 1 0 0 0 0 0 1 0 0 0 0 0 1 1 1 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 0 1 0 0
##    [852] 1 0 1 0 0 0 0 0 1 1 0 0 1 1 1 1 0 1 0 0 0 1 0 0 0 0 0 1 0 1 0 0 1 0 1 1
##    [889] 1 0 0 1 1 1 1 1 1 0 1 1 0 0 0 1 1 1 1 0 0 0 0 1 0 0 0 0 0 1 1 1 0 0 1 0 1
##    [926] 1 0 1 1 0 0 0 1 1 1 0 0 1 0 0 0 0 0 0 1 1 1 0 1 1 0 0 0 1 0 0 1 0 1 0 1
##    [963] 1 1 1 1 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 1 0 1 1 0 0 0 1 0 0 0 1 0 1 0 1
##   [1000] 0 0 0 1 0 1 1 1 0 1 0 1 0 0 0 0 0 0 1 0 0 0 0 0 0 0 1 0 0 0 1 0 0 1 1 0
##   [1037] 1 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 0 0 0 0 0 1 0 0 1 0 0 0 0 1 1 1
##   [1074] 1 1 0 1 1 0 0 0 0 1 0 1 0 0 0 0 1 0 0 0 0 0 1 0 0 1 0 1 0 0 1 0 1 0 0 1
##   [1111] 1 0 0 0 0 0 1 0 0 0 1 1 1 0 1 0 0 0 0 0 1 1 0 0 1 0 0 0 0 1 1 0 0 0 0 0
##   [1148] 0 0 0 1 0 1 0 1 0 0 0 0 1 1 0 0 0 1 1 1 0 1 0 1 0 1 0 1 0 1 1 1 0 1 0 1
##   [1185] 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 1 1 0 1 0 0 0 0 0 0 1 1 0 0 1 0
##   [1222] 0 0 0 1 0 0 0 0 0 0 0 1 0 0 0 0 0 1 1 0 1 0 0 0 0 0 1 0 1 0 0 0 1 0 0 1
##   [1259] 0 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 1 1 1 0 0 0 0 1 1 1 1 0 0 1 1 0 1 1 0
##   [1296] 1 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 1 0 1 1 1 1 0 0 0 1 0 0 1 1 1 0 0 0 0
##   [1333] 1 1 1 0 1 1 0 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 0 1 0 1 1 0 1 0 1 0 1 0
##   [1370] 0 0 0 0 1 1 1 0 0 1 0 0 0 1 1 1 1 0 1 1 0 0 1 0 1 0 0 0 0 0 0 0 0 0 0 0
##   [1407] 1 0 1 1 0 0 1 1 0 1 1 0 0 0 0 0 1 0 1 0 0 1 0 0 0 1 0 1 0 1 0 0 0 0 1 0
##   [1444] 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 0 0 0 1 0 0 0 0 1 0 0 1 0 0 0 0 0 0 0 0 0
##   [1481] 1 0 0 1 1 0 0 0 1 0 1 0 0 0 0 1 1 0 0 0 0 1 0 0 1 0 0 0 0 1 1 0 1 1 0 0 1
##   [1518] 0 0 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0 0 1 1 0 0 1 0 0 1 0 0 1 1 1 0 0 0 0 0 1
##   [1555] 0 1 1 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 1 0 1 0 0 1 1 1 0 0 1 0 0
##   [1592] 1 0 1 0 1 1 0 0 0 0 0 0 1 0 0 0 1 1 0 0 0 0 1 0 0 0 1 0 0 1 0 1 0 1 0 0 0

```

```
## [1629] 1 0 0 0 0 0 0 0 0 0 0 1 0 1 0 0 1 0 1 0 1 0 0 1 0 1 0 0 0 1 1 0 0 1 0 0 0
## [1666] 0 0 0 0 0 0 1 0 0 0 1 1 0 0 1 1 0 1 0 0 0 1 1 1 0 1 1 1 0 1 1 1 0 0 0 1 1
## [1703] 0 0 0 1 0 1 0 0 0 0 0 0 1 0 1 0 0 1 1 0 0 0 0 0 1 0 0 1 0 0 0 1 0 0 1 0 0
## [1740] 0 0 1 0 0 0 1 0 0 0 0 0 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 0 0 1
## [1777] 0 0 0 0 0 1 0 0 0 1 1 0 0 0 1 0 0 0 0 1 0 0 0 0 1 0 0 0 1 0 0 1 1 1 0 0 0
## [1814] 0 1 0 0 0 1 0 1 0 0 1 1 0 0 0 0 0 0 0 1 0 0 0 0 1 0 1 0 0 0 1 0 1 0 0 0 0
## [1851] 1 0 1 0 0 0 1 0 1 0 1 0 0 0 1 0 1 0 1 0 0 0 0 1 0 1 1 0 1 0 1 0 0 1 0 0 0
## [1888] 1 0 1 0 1 0 0 0 1 1 1 1 0 0 0 0 0 0 0 1 1 0 0 1 0 0 0 0 1 0 0 0 0 0 1 1 1
## [1925] 0 0 0 0 0 0 1 1 1 0 0 1 0 0 0 0 0 0 0 0 1 0 1 1 1 1 1 0 0 1 0 1 1 1 0 0 1
## [1962] 1 0 0 0 1 1 1 1 1 0 0 0 0 0 1 1 0 1 0 1 0 1 0 1 0 0 0 1 0 0 1 1 0 1 0 0 0
## [1999] 1 0 0 0 0 1 1 0 1 1 0 0 1 0 0 0 1 1 0 1 0 1 0 1 1 0 0 0 1 0 0 1 0 0 1 0 1
## [2036] 1 1 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 1 0 0 0 0 1 1 1 1 0 1 1 0 0 1 0 0 0 0 0
## [2073] 0 0 0 1 1 0 0 0 0 0 0 1 1 0 1 0 0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 0 0 0
## [2110] 1 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 1 0 1 0 1 0
## Levels: 0 1
```

```
# Making the Confusion Matrix
```

```
cm = table(test_set[, 17], y_pred)
cm
```

```
##      y_pred
##      0      1
## 0 1107  133
## 1   282  610
```

```
print("=====KNN=====")
```

```
## [1] "=====KNN====="
```

```
library(ggplot2)
library(lattice)
library(caret)
confusionMatrix(cm)
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##      y_pred
##      0      1
## 0 1107  133
## 1   282  610
```

```
##
```

```
##              Accuracy : 0.8053
##              95% CI   : (0.7879, 0.822)
##    No Information Rate : 0.6515
##    P-Value [Acc > NIR] : < 2.2e-16
```

```
##
```

```
##              Kappa : 0.5904
```

```
##
```

```
## Mcnemar's Test P-Value : 3.729e-13
```

```
##
```

```
##              Sensitivity : 0.7970
```

```
##              Specificity : 0.8210
```

```
##              Pos Pred Value : 0.8927
```

```
##              Neg Pred Value : 0.6839
```

```
##              Prevalence : 0.6515
```

```
##              Detection Rate : 0.5192
```

	Training error	CV error	Public Leaderboard error (if available)
kNN	...	...	...
Ridge	...	...	...
lasso	...	...	...
ElasticNet	...	...	...
random forest	...	...	
SVM	...	...	
LDA	...	...	
QDA	...	...	
C5.0	...	...	

Table 3: Training and CV error of the different models.

```
## Detection Prevalence : 0.5816
## Balanced Accuracy : 0.8090
##
## 'Positive' Class : 0
##
```

## 6 Some general comments

Here are some comments that apply to many of the intermediate reports.

- Write your team name on the front page.
- Write you own original report, do not follow step by step the exercise but rather think about a useful structure for your own report.
- Use plots and figures, but only those that contain relevant information.
- In R plots, make sure that there are labels on each axis, that they are readable, that there is a caption that says what the plot or figure shows, the size of the points/lines is appropriate, the form of the plot is as you want it (e.g., squared), etc.
- Use tables, again be careful to describe what the table shows in a caption.
- Avoid lengthy R output of model fits.
- Explain what you do: if you use a model, give a brief description in terms of a formula. You don't have to reproduce what we did in lecture, but the report should be readable on its own and be consistently written and structured.
- In the same vein as the previous comment, if you use notation like AIC, explain what this is and why you use it.
- The response is not required to be normal in a linear model (or any other method), it is only assumed to be normal conditional on the predictor values, that is, the residuals should be normal. Slight violation of this does not mean that the model is useless for prediction. Test its performance with CV.
- Careful with excluding predictors, even if there is high correlation, there might be additional information. Test with CV.
- It is always good to know the difficulty of the problem. One can assess this by starting with simple benchmark models like the overall mean, kNN, LM, etc. In the report you should try these models and report their errors, this gives you also a feeling about what type of model performs well on this data. You don't have to report the errors 20 different LMs with interactions, just choose the relevant models (e.g., the best LM with interactions).

- If you use CV (and you should), describe what you do exactly. Make sure to use the same CV for all models to make them comparable.
- Try to use Latex or RMarkdown, it looks better!

## 7 Some tests

Hello test