

Bank_Data_Factor_Analysis.R

deepti

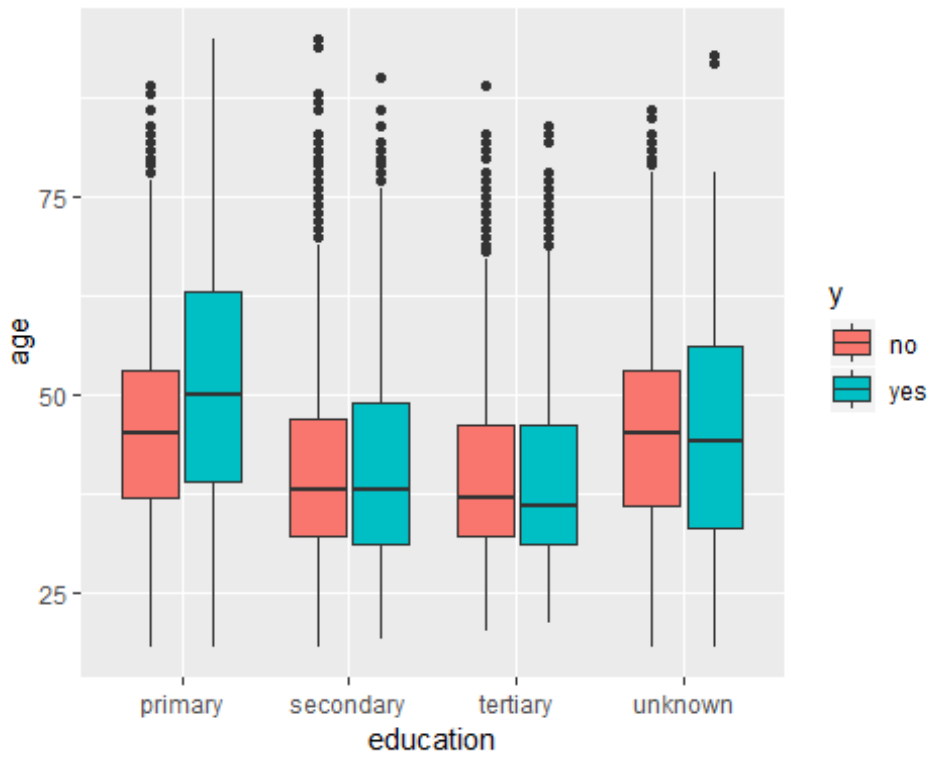
Thu Mar 28 21:52:18 2019

```
bank <- read.csv("~/Spring 19 Sem/Multi Analysis/bank-full.csv", sep=";")
str(bank)

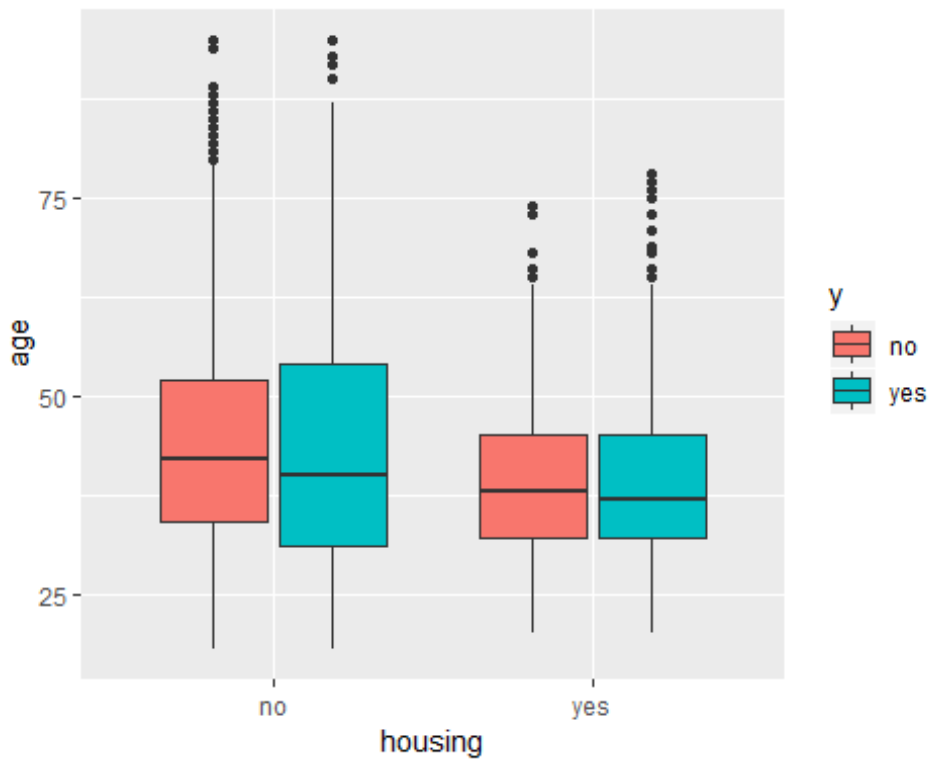
## 'data.frame':    45211 obs. of  17 variables:
## $ age          : int  58 44 33 47 33 35 28 42 58 43 ...
## $ job          : Factor w/ 12 levels "admin.,""blue-collar",...: 5 10 3 2 12 5
5 3 6 10 ...
## $ marital      : Factor w/ 3 levels "divorced","married",...: 2 3 2 2 3 2 3 1
2 3 ...
## $ education: Factor w/ 4 levels "primary","secondary",...: 3 2 2 4 4 3 3 3
1 2 ...
## $ default      : Factor w/ 2 levels "no","yes": 1 1 1 1 1 1 1 2 1 1 ...
## $ balance      : int  2143 29 2 1506 1 231 447 2 121 593 ...
## $ housing      : Factor w/ 2 levels "no","yes": 2 2 2 2 1 2 2 2 2 2 ...
## $ loan         : Factor w/ 2 levels "no","yes": 1 1 2 1 1 1 2 1 1 1 ...
## $ contact      : Factor w/ 3 levels "cellular","telephone",...: 3 3 3 3 3 3 3
3 3 3 ...
## $ day          : int  5 5 5 5 5 5 5 5 5 5 ...
## $ month        : Factor w/ 12 levels "apr","aug","dec",...: 9 9 9 9 9 9 9 9 9
9 ...
## $ duration     : int  261 151 76 92 198 139 217 380 50 55 ...
## $ campaign     : int  1 1 1 1 1 1 1 1 1 1 ...
## $ pdays        : int  -1 -1 -1 -1 -1 -1 -1 -1 -1 -1 ...
## $ previous     : int  0 0 0 0 0 0 0 0 0 0 ...
## $ poutcome     : Factor w/ 4 levels "failure","other",...: 4 4 4 4 4 4 4 4 4 4
...
## $ y            : Factor w/ 2 levels "no","yes": 1 1 1 1 1 1 1 1 1 1 ...

attach(bank)

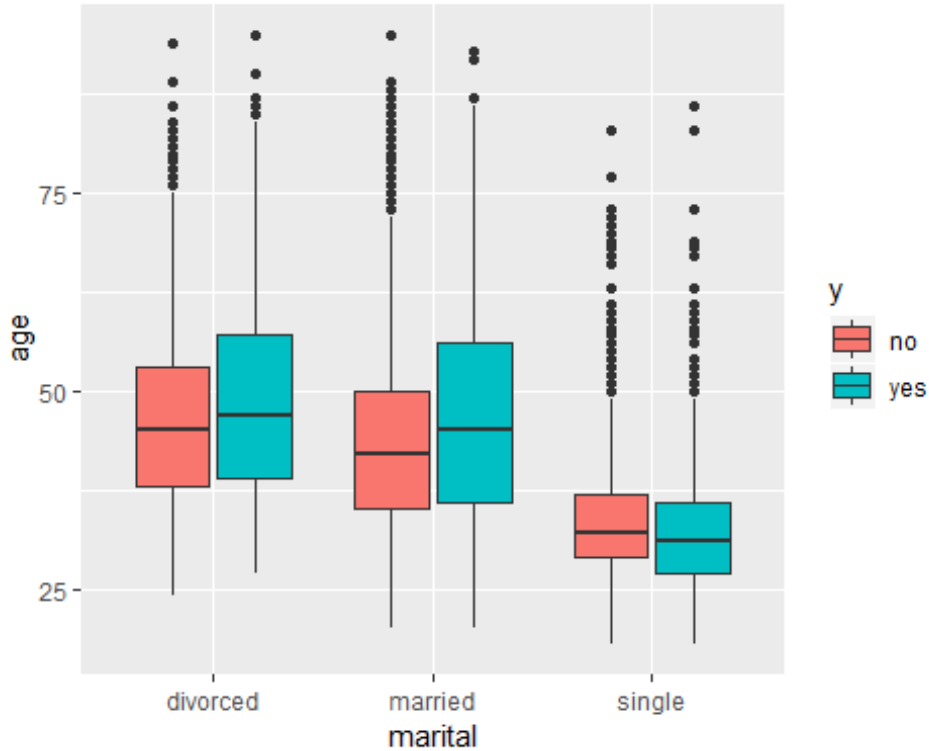
library(ggplot2)
ggplot(bank, aes(y=age,x=education, fill=y)) + geom_boxplot()
```



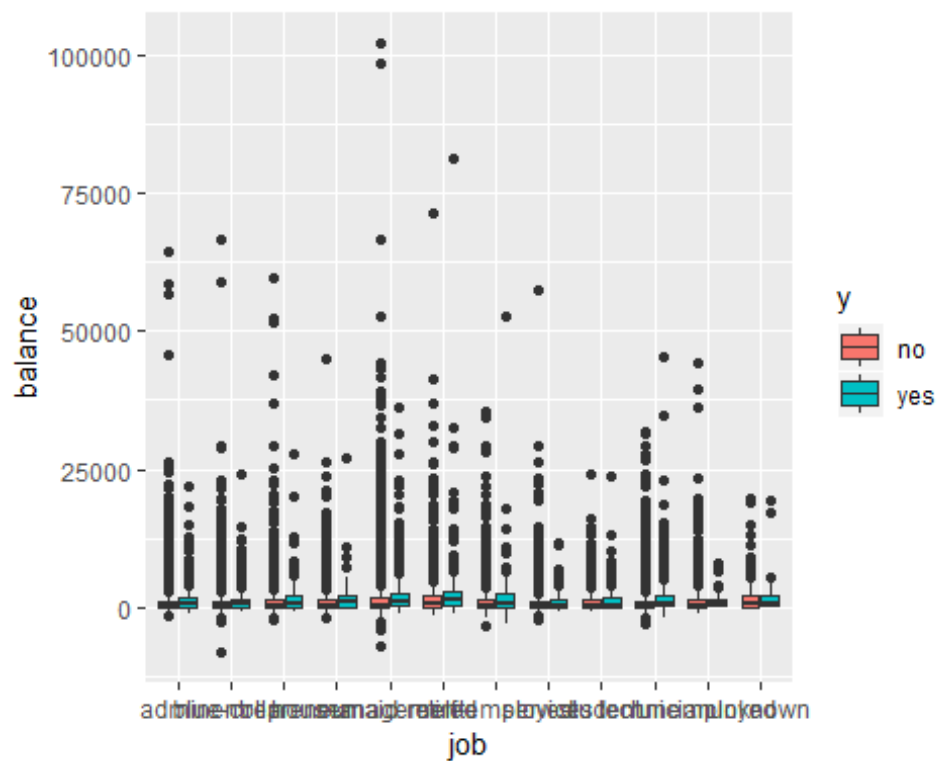
```
ggplot(bank, aes(y=age,x=housing, fill=y)) + geom_boxplot()
```



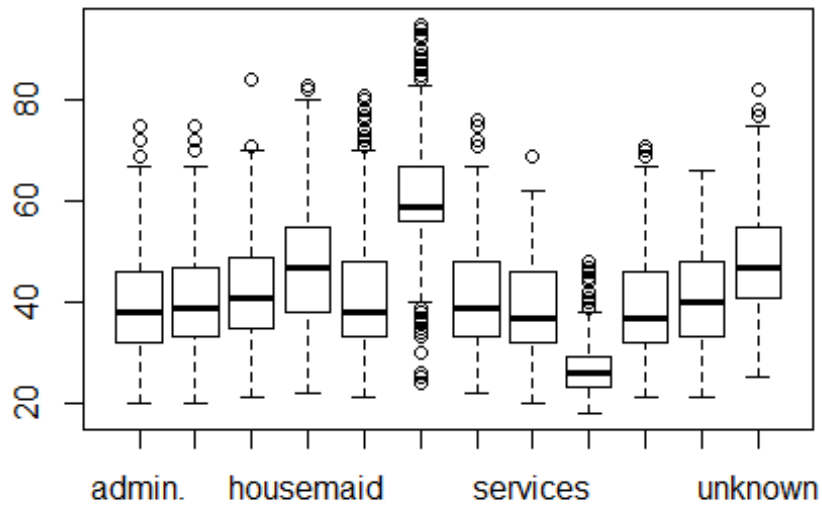
```
ggplot(bank, aes(y=age,x=marital, fill=y)) + geom_boxplot()
```



```
ggplot(bank, aes(y=balance,x=job, fill=y)) + geom_boxplot()
```



```
boxplot(age ~ job)
```



```
unique(bank$default)
```

```
## [1] no yes  
## Levels: no yes
```

```
default_dummy=ifelse(bank$default=='yes',1,0)
```

```
unique(bank$housing)
```

```
## [1] yes no  
## Levels: no yes
```

```
housing_dummy=ifelse(bank$housing=='yes',1,0)
```

```
unique(bank$loan)
```

```
## [1] no yes  
## Levels: no yes
```

```
loan_dummy=ifelse(bank$loan=='yes',1,0)
```

```
unique(bank$contact)
```

```
## [1] unknown cellular telephone  
## Levels: cellular telephone unknown
```

```

cell_dummy=ifelse(bank$contact=='cellular',1,0)

unique(bank$marital)

## [1] married single divorced
## Levels: divorced married single

married_dummy=ifelse(bank$marital=='married',1,0)
divorced_dummy=ifelse(bank$marital=='divorced',1,0)

admin_dummy=ifelse(bank$job== 'admin.',1,0)
bluecollar_dummy=ifelse(bank$job== 'blue-collar',1,0)
technician_dummy=ifelse(bank$job== 'technician',1,0)
services_dummy=ifelse(bank$job=='services',1,0)
management_dummy=ifelse(bank$job=='management',1,0)
retired_dummy=ifelse(bank$job=='retired',1,0)
entrepreneur_dummy=ifelse(bank$job=='entrepreneur',1,0)
selfemployed_dummy=ifelse(bank$job=='self-employed',1,0)
housemaid_dummy=ifelse(bank$job=='housemaid',1,0)
unemployed_dummy=ifelse(bank$job=='unemployed',1,0)
student_dummy=ifelse(bank$job=='student',1,0)

bank1=data.frame(age,balance,day,duration,campaign,pdays,previous,default_dum
my,housing_dummy,loan_dummy,cell_dummy,married_dummy,divorced_dummy,

admin_dummy,bluecollar_dummy,technician_dummy,services_dummy,management_dummy
,retired_dummy,

entrepreneur_dummy,selfemployed_dummy,housemaid_dummy,unemployed_dummy,student_dummy)

# Computing Correlation Matrix
corrm.job <- cor(bank1)
corrm.job

##              age      balance      day      duration
## age          1.0000000000  0.097782739 -0.009120046 -0.004648428
## balance      0.0977827394  1.000000000  0.004502585  0.021560380
## day         -0.0091200456  0.004502585  1.000000000 -0.030206341
## duration    -0.0046484285  0.021560380 -0.030206341  1.000000000
## campaign     0.0047603118 -0.014578279  0.162490216 -0.084569503
## pdays       -0.0237580141  0.003435322 -0.093044074 -0.001564770
## previous     0.0012883192  0.016673637 -0.051710497  0.001203057
## default_dummy -0.0178793036 -0.066745057  0.009423899 -0.010021461
## housing_dummy -0.1855130815 -0.068768316 -0.027981649  0.005075449
## loan_dummy   -0.0156552727 -0.084350246  0.011370158 -0.012411972
## cell_dummy   -0.0683752110  0.015879610  0.020185831  0.025542949
## married_dummy  0.2862568801  0.025708074  0.007102369 -0.022682929
## divorced_dummy 0.1646742841 -0.021731604 -0.000433111  0.006099890
## admin_dummy   -0.0557168481 -0.026725827 -0.010454887 -0.015721865

```

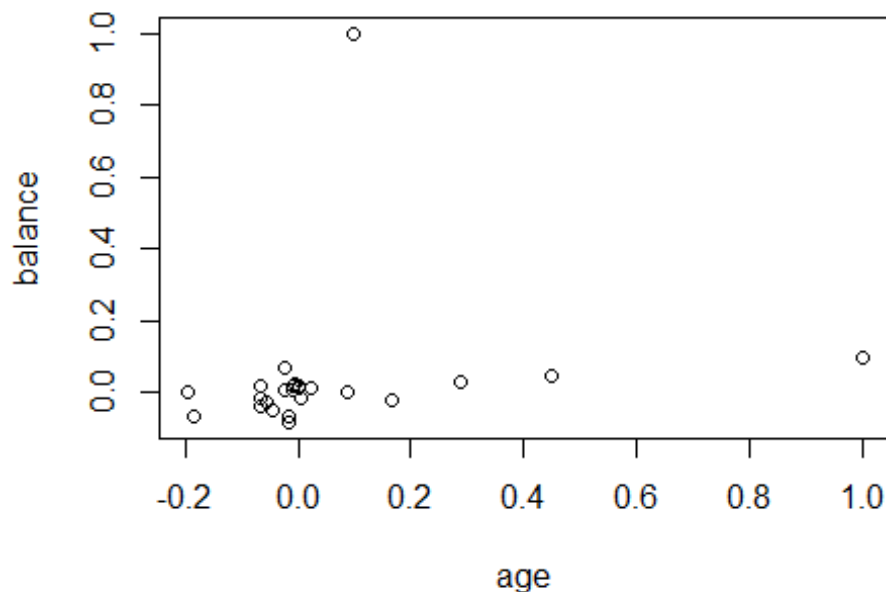
## bluecollar_dummy	-0.0440021457	-0.048756856	-0.022898115	0.009636854
## technician_dummy	-0.0686316215	-0.016183263	0.032517190	-0.009176081
## services_dummy	-0.0657801637	-0.038150675	-0.006531311	0.001427402
## management_dummy	-0.0235714033	0.067797227	0.019020516	-0.008322998
## retired_dummy	0.4473782669	0.046900051	-0.010122986	0.026032085
## entrepreneur_dummy	0.0217918185	0.009642377	-0.002311926	-0.001327465
## selfemployed_dummy	-0.0080927264	0.017850395	0.005061864	0.007382572
## housemaid_dummy	0.0866500450	0.001661416	0.003954916	-0.008045553
## unemployed_dummy	0.0004123369	0.009022792	-0.006382325	0.020322299
## student_dummy	-0.1973019012	0.001232856	-0.015894080	-0.006503553
##	campaign	pdays	previous	default_dummy
## age	0.004760312	-0.023758014	0.001288319	-1.787930e-02
## balance	-0.014578279	0.003435322	0.016673637	-6.674506e-02
## day	0.162490216	-0.093044074	-0.051710497	9.423899e-03
## duration	-0.084569503	-0.001564770	0.001203057	-1.002146e-02
## campaign	1.000000000	-0.088627668	-0.032855290	1.682153e-02
## pdays	-0.088627668	1.000000000	0.454819635	-2.997936e-02
## previous	-0.032855290	0.454819635	1.000000000	-1.832940e-02
## default_dummy	0.016821531	-0.029979361	-0.018329405	1.000000e+00
## housing_dummy	-0.023598707	0.124178400	0.037076150	-6.025218e-03
## loan_dummy	0.009979846	-0.022753639	-0.011043488	7.723424e-02
## cell_dummy	-0.032270965	0.225996328	0.131716708	-1.006107e-02
## married_dummy	0.031371290	-0.027600380	-0.012690602	-1.445963e-02
## divorced_dummy	-0.015483915	0.002896612	-0.004534419	1.777686e-02
## admin_dummy	-0.021868078	0.027496834	0.014245294	-1.003715e-02
## bluecollar_dummy	0.008986250	0.020063311	-0.017095359	1.034087e-02
## technician_dummy	0.020739365	-0.013477571	-0.001122739	-3.089271e-03
## services_dummy	-0.004671344	0.005709790	-0.010925770	6.765902e-05
## management_dummy	0.016686303	-0.007871366	0.019578719	-2.654893e-03
## retired_dummy	-0.030913063	-0.006314946	0.005818128	-1.129037e-02
## entrepreneur_dummy	0.002128448	-0.014203221	-0.008180632	2.628100e-02
## selfemployed_dummy	0.005479290	-0.010355532	-0.002370967	4.107473e-03
## housemaid_dummy	0.003096636	-0.031349724	-0.015204210	-3.592806e-04
## unemployed_dummy	-0.018447859	-0.010410966	-0.008503909	6.470248e-03
## student_dummy	-0.021813247	0.024485938	0.023555842	-1.622245e-02
##	housing_dummy	loan_dummy	cell_dummy	married_dummy
## age	-0.185513082	-0.015655273	-0.0683752110	0.286256880
## balance	-0.068768316	-0.084350246	0.0158796096	0.025708074
## day	-0.027981649	0.011370158	0.0201858307	0.007102369
## duration	0.005075449	-0.012411972	0.0255429485	-0.022682929
## campaign	-0.023598707	0.009979846	-0.0322709652	0.031371290
## pdays	0.124178400	-0.022753639	0.2259963279	-0.027600380
## previous	0.037076150	-0.011043488	0.1317167080	-0.012690602
## default_dummy	-0.006025218	0.077234241	-0.0100610716	-0.014459629
## housing_dummy	1.000000000	0.041322866	-0.1560708266	0.017680694
## loan_dummy	0.041322866	1.000000000	0.0136038591	0.036905561
## cell_dummy	-0.156070827	0.013603859	1.0000000000	-0.043477419
## married_dummy	0.017680694	0.036905561	-0.0434774188	1.000000000
## divorced_dummy	0.001778034	0.016376783	-0.0067862807	-0.443647677
## admin_dummy	0.043046134	0.030781333	0.0006595637	-0.059572424

## bluecollar_dummy	0.177474869	0.018291405	-0.1293159114	0.122033433
## technician_dummy	-0.012822887	0.014798107	0.0569820979	-0.062949878
## services_dummy	0.070435297	0.035574867	-0.0308926754	-0.014615340
## management_dummy	-0.063382009	-0.038904191	0.1018532420	-0.032559955
## retired_dummy	-0.156632670	-0.014861001	-0.0204843406	0.076287703
## entrepreneur_dummy	0.010600051	0.039808166	-0.0023860562	0.044318525
## selfemployed_dummy	-0.027318952	-0.007881820	0.0154401257	0.010471978
## housemaid_dummy	-0.079380025	-0.017234414	-0.0156494875	0.045816483
## unemployed_dummy	-0.048228784	-0.035960853	0.0130066836	-0.014400748
## student_dummy	-0.085065505	-0.058503161	0.0290488286	-0.161871017
##	divorced_dummy	admin_dummy	bluecollar_dummy	
## age	0.164674284	-0.0557168481	-0.044002146	
## balance	-0.021731604	-0.0267258266	-0.048756856	
## day	-0.000433111	-0.0104548871	-0.022898115	
## duration	0.006099890	-0.0157218649	0.009636854	
## campaign	-0.015483915	-0.0218680777	0.008986250	
## pdays	0.002896612	0.0274968340	0.020063311	
## previous	-0.004534419	0.0142452936	-0.017095359	
## default_dummy	0.017776857	-0.0100371471	0.010340865	
## housing_dummy	0.001778034	0.0430461339	0.177474869	
## loan_dummy	0.016376783	0.0307813332	0.018291405	
## cell_dummy	-0.006786281	0.0006595637	-0.129315911	
## married_dummy	-0.443647677	-0.0595724241	0.122033433	
## divorced_dummy	1.000000000	0.0336242881	-0.062517892	
## admin_dummy	0.033624288	1.0000000000	-0.188215593	
## bluecollar_dummy	-0.062517892	-0.1882155926	1.0000000000	
## technician_dummy	0.009273970	-0.1615052103	-0.235375724	
## services_dummy	0.016929695	-0.1143088549	-0.166592331	
## management_dummy	0.003698621	-0.1848347587	-0.269375923	
## retired_dummy	0.052180279	-0.0825110340	-0.120250575	
## entrepreneur_dummy	0.003007167	-0.0662728696	-0.096585272	
## selfemployed_dummy	-0.015796245	-0.0683642005	-0.099633152	
## housemaid_dummy	0.017473252	-0.0603486981	-0.087951457	
## unemployed_dummy	0.008668957	-0.0619071220	-0.090222684	
## student_dummy	-0.049597256	-0.0523084894	-0.076233754	
##	technician_dummy	services_dummy	management_dummy	
## age	-0.068631622	-6.578016e-02	-0.023571403	
## balance	-0.016183263	-3.815068e-02	0.067797227	
## day	0.032517190	-6.531311e-03	0.019020516	
## duration	-0.009176081	1.427402e-03	-0.008322998	
## campaign	0.020739365	-4.671344e-03	0.016686303	
## pdays	-0.013477571	5.709790e-03	-0.007871366	
## previous	-0.001122739	-1.092577e-02	0.019578719	
## default_dummy	-0.003089271	6.765902e-05	-0.002654893	
## housing_dummy	-0.012822887	7.043530e-02	-0.063382009	
## loan_dummy	0.014798107	3.557487e-02	-0.038904191	
## cell_dummy	0.056982098	-3.089268e-02	0.101853242	
## married_dummy	-0.062949878	-1.461534e-02	-0.032559955	
## divorced_dummy	0.009273970	1.692970e-02	0.003698621	
## admin_dummy	-0.161505210	-1.143089e-01	-0.184834759	

## bluecollar_dummy	-0.235375724	-1.665923e-01	-0.269375923
## technician_dummy	1.000000000	-1.429506e-01	-0.231147773
## services_dummy	-0.142950587	1.000000e+00	-0.163599906
## management_dummy	-0.231147773	-1.635999e-01	1.000000000
## retired_dummy	-0.103185363	-7.303170e-02	-0.118090566
## entrepreneur_dummy	-0.082878493	-5.865907e-02	-0.094850353
## selfemployed_dummy	-0.085493837	-6.051014e-02	-0.097843485
## housemaid_dummy	-0.075469935	-5.341550e-02	-0.086371623
## unemployed_dummy	-0.077418844	-5.479489e-02	-0.088602054
## student_dummy	-0.065415136	-4.629900e-02	-0.074864401
##	retired_dummy	entrepreneur_dummy	selfemployed_dummy
## age	0.447378267	0.021791819	-0.008092726
## balance	0.046900051	0.009642377	0.017850395
## day	-0.010122986	-0.002311926	0.005061864
## duration	0.026032085	-0.001327465	0.007382572
## campaign	-0.030913063	0.002128448	0.005479290
## pdays	-0.006314946	-0.014203221	-0.010355532
## previous	0.005818128	-0.008180632	-0.002370967
## default_dummy	-0.011290368	0.026280997	0.004107473
## housing_dummy	-0.156632670	0.010600051	-0.027318952
## loan_dummy	-0.014861001	0.039808166	-0.007881820
## cell_dummy	-0.020484341	-0.002386056	0.015440126
## married_dummy	0.076287703	0.044318525	0.010471978
## divorced_dummy	0.052180279	0.003007167	-0.015796245
## admin_dummy	-0.082511034	-0.066272870	-0.068364200
## bluecollar_dummy	-0.120250575	-0.096585272	-0.099633152
## technician_dummy	-0.103185363	-0.082878493	-0.085493837
## services_dummy	-0.073031704	-0.058659071	-0.060510138
## management_dummy	-0.118090566	-0.094850353	-0.097843485
## retired_dummy	1.000000000	-0.042341607	-0.043677754
## entrepreneur_dummy	-0.042341607	1.000000000	-0.035081976
## selfemployed_dummy	-0.043677754	-0.035081976	1.000000000
## housemaid_dummy	-0.038556665	-0.030968717	-0.031945977
## unemployed_dummy	-0.039552339	-0.031768442	-0.032770939
## student_dummy	-0.033419792	-0.026842780	-0.027689840
##	housemaid_dummy	unemployed_dummy	student_dummy
## age	0.0866500450	0.0004123369	-0.197301901
## balance	0.0016614156	0.0090227918	0.001232856
## day	0.0039549157	-0.0063823246	-0.015894080
## duration	-0.0080455532	0.0203222992	-0.006503553
## campaign	0.0030966362	-0.0184478589	-0.021813247
## pdays	-0.0313497236	-0.0104109656	0.024485938
## previous	-0.0152042099	-0.0085039090	0.023555842
## default_dummy	-0.0003592806	0.0064702478	-0.016222451
## housing_dummy	-0.0793800250	-0.0482287840	-0.085065505
## loan_dummy	-0.0172344136	-0.0359608525	-0.058503161
## cell_dummy	-0.0156494875	0.0130066836	0.029048829
## married_dummy	0.0458164829	-0.0144007479	-0.161871017
## divorced_dummy	0.0174732524	0.0086689569	-0.049597256
## admin_dummy	-0.0603486981	-0.0619071220	-0.052308489


```
## bluecollar_dummy    -0.0879514566    -0.0902226845    -0.076233754
## technician_dummy   -0.0754699347    -0.0774188442    -0.065415136
## services_dummy      -0.0534155016    -0.0547948850    -0.046298997
## management_dummy    -0.0863716232    -0.0886020541    -0.074864401
## retired_dummy       -0.0385566654    -0.0395523394    -0.033419792
## entrepreneur_dummy  -0.0309687170    -0.0317684424    -0.026842780
## selfemployed_dummy  -0.0319459771    -0.0327709389    -0.027689840
## housemaid_dummy     1.0000000000    -0.0289286423    -0.024443287
## unemployed_dummy    -0.0289286423    1.0000000000    -0.025074502
## student_dummy       -0.0244432874    -0.0250745024    1.000000000
```

```
plot(corrmat.job)
```



```
bank_pca <- prcomp(bank1, scale=TRUE)
bank_pca
```

```
## Standard deviations (1, ..., p=24):
## [1] 1.33844965 1.28503448 1.24822205 1.17257046 1.11445100 1.09587255
## [7] 1.07326056 1.05901699 1.04666001 1.04300561 1.02140057 1.01735820
## [13] 1.01416721 1.01229356 0.98720911 0.95730835 0.94523110 0.91925677
## [19] 0.90856408 0.83749497 0.76847418 0.70991612 0.57851915 0.08595291
##
## Rotation (n x k) = (24 x 24):
##               PC1          PC2          PC3          PC4
## age          0.569318294 -0.20981495  0.11519363  0.220812369
## balance      0.136431648 -0.17156951  0.02405785 -0.163302019
## day          0.059418593  0.07154041 -0.19566750 -0.195731497
```

## duration	-0.007808696	-0.05044539	0.03322156	0.077099400
## campaign	0.071690584	0.12991463	-0.14965979	-0.192439063
## pdays	-0.267207562	-0.32389991	0.48657173	0.075716057
## previous	-0.207016894	-0.33203403	0.43859732	0.031886666
## default_dummy	-0.011688999	0.07355946	-0.06880501	0.059004928
## housing_dummy	-0.256101191	0.31697015	0.22466536	0.206205407
## loan_dummy	-0.018056549	0.12223324	0.01063137	0.121024200
## cell_dummy	-0.155137611	-0.37743547	0.08347566	-0.248250473
## married_dummy	0.377325484	0.22379032	0.40990646	-0.292625174
## divorced_dummy	-0.059777795	-0.24896983	-0.31775550	0.525594559
## admin_dummy	-0.118692793	-0.02269682	-0.02667375	0.214602390
## bluecollar_dummy	-0.005484778	0.41751730	0.32228852	0.145468149
## technician_dummy	-0.097685059	-0.04674918	-0.16057401	-0.051013476
## services_dummy	-0.084735874	0.08390258	-0.02225662	0.143199790
## management_dummy	-0.014553299	-0.22637577	-0.10108961	-0.423658222
## retired_dummy	0.448368496	-0.25464810	0.06962613	0.248685264
## entrepreneur_dummy	0.046291161	0.02949425	0.01538236	0.008088858
## selfemployed_dummy	0.018603913	-0.01428582	-0.01226736	-0.074764293
## housemaid_dummy	0.135388675	-0.03314405	-0.04067568	-0.005552555
## unemployed_dummy	0.015204401	-0.05129622	-0.04771282	-0.004344841
## student_dummy	-0.212304260	-0.08086513	-0.11711951	-0.122560143
##	PC5	PC6	PC7	PC8
## age	0.05045844	-0.090007424	0.07343767	0.0029022871
## balance	-0.20316286	0.108831230	0.08869295	0.0369876194
## day	0.23299774	-0.204016399	0.17551554	0.1838922279
## duration	-0.16835993	0.199240499	-0.05544368	-0.2600908646
## campaign	0.23357320	-0.259229849	0.22022006	0.2233641536
## pdays	0.10406098	-0.071142929	0.07473538	0.0280133584
## previous	0.13165009	-0.093079877	0.09362569	0.0569267720
## default_dummy	0.14524402	-0.199070450	-0.08027030	-0.1205427310
## housing_dummy	-0.03576172	-0.180480835	0.10334660	-0.0613144125
## loan_dummy	0.30823708	-0.282779230	-0.20115497	-0.1188024046
## cell_dummy	0.19185804	-0.030387100	-0.06894623	-0.0314369971
## married_dummy	0.13525628	0.001571138	-0.17662366	-0.0145079214
## divorced_dummy	-0.07128989	-0.210401979	0.25814558	-0.0265230546
## admin_dummy	0.03110704	-0.107824160	-0.44336056	0.6689262746
## bluecollar_dummy	-0.20852017	0.050528440	0.43721502	0.0917732995
## technician_dummy	0.61663938	0.341504993	0.31142644	-0.1607766706
## services_dummy	0.03130057	-0.110199160	-0.37625264	-0.5135887546
## management_dummy	-0.36771755	-0.508552464	0.12128887	-0.1541152341
## retired_dummy	0.02306536	0.075265038	0.03780626	0.0002201704
## entrepreneur_dummy	0.09763562	-0.078029069	-0.21313877	-0.0920294502
## selfemployed_dummy	0.01471725	0.105492017	-0.14845733	0.0418563401
## housemaid_dummy	0.01626816	0.080068700	-0.09485817	0.1016528877
## unemployed_dummy	-0.09127054	0.192415138	-0.08805854	-0.0423535915
## student_dummy	-0.18301211	0.390586165	-0.05655534	0.1316410580
##	PC9	PC10	PC11	PC12
## age	-0.02055961	0.01899410	0.003017453	0.0061180216
## balance	-0.06349185	0.28109715	0.186985066	-0.1570666075
## day	-0.25370231	-0.05582818	0.187984088	0.0404964573

## duration	0.27380295	0.09689058	0.172574465	0.0702481546
## campaign	-0.33309209	-0.11620390	0.123069087	0.0084341860
## pdays	-0.10560222	-0.05965036	0.025312617	0.0327996841
## previous	-0.14273017	-0.09073198	0.031277498	0.0259227801
## default_dummy	0.26788782	-0.46161857	-0.000198242	0.0905748007
## housing_dummy	-0.01429705	0.26378799	0.089905643	-0.0221716052
## loan_dummy	0.24570045	-0.17099519	-0.044743674	-0.0592999846
## cell_dummy	0.10782075	-0.12818491	0.027332335	0.0805791579
## married_dummy	0.02930515	0.08186986	-0.049544742	0.0483345935
## divorced_dummy	-0.01103252	-0.04156962	0.101297102	0.0367078636
## admin_dummy	0.12267870	0.29063350	-0.043401009	0.0214157028
## bluecollar_dummy	0.05352306	-0.24481249	0.010733462	0.0154211106
## technician_dummy	0.17669076	0.31086242	-0.095081041	-0.0199198864
## services_dummy	-0.59822048	0.11051397	-0.048043636	0.0163644902
## management_dummy	0.18907075	0.11586201	-0.121895314	0.0008360632
## retired_dummy	-0.08515263	-0.07915230	-0.047755350	-0.2067772045
## entrepreneur_dummy	0.22329742	-0.20765619	0.137250169	-0.4563233943
## selfemployed_dummy	-0.07000420	-0.11898191	0.761619747	-0.0437733631
## housemaid_dummy	-0.13363741	-0.19801333	-0.354153598	0.4021783703
## unemployed_dummy	0.07287735	-0.12818443	0.251120230	0.6407257567
## student_dummy	-0.20214602	-0.40023847	-0.204254071	-0.3434643188
##	PC13	PC14	PC15	PC16
## age	0.00608206	-0.004366293	0.0337528061	-0.006569485
## balance	0.22759641	-0.105046637	-0.0520671815	0.371870157
## day	0.08560075	0.152146462	-0.4412207944	0.058281117
## duration	-0.09912347	0.045680280	-0.7425113763	0.174578677
## campaign	0.11270968	0.081091964	-0.0904947294	0.062386121
## pdays	0.04785529	-0.019203409	-0.0006823068	0.043047490
## previous	0.06629664	-0.029377127	0.0032209294	0.122485125
## default_dummy	-0.13009310	0.078299387	0.1514064688	0.737797404
## housing_dummy	0.10524623	-0.078688089	0.0381481909	0.110299856
## loan_dummy	-0.16464085	0.062435569	-0.1247403804	-0.344748580
## cell_dummy	-0.08273178	0.075980512	-0.2120860907	-0.227977287
## married_dummy	-0.00770238	0.024366132	0.0173287855	0.012932230
## divorced_dummy	0.07179413	-0.132889375	-0.0155413006	-0.088530731
## admin_dummy	-0.04896376	0.123996856	-0.0869130468	0.127403587
## bluecollar_dummy	-0.06833214	0.046445448	-0.1374761952	-0.115721043
## technician_dummy	-0.01957358	-0.046238958	0.1130159712	0.092217511
## services_dummy	-0.01559822	0.099301632	-0.0567885916	0.070887107
## management_dummy	-0.09814717	-0.004809872	0.0920088209	-0.030629021
## retired_dummy	-0.14461331	0.276117409	-0.0039226996	0.011479493
## entrepreneur_dummy	0.68685598	-0.206269448	-0.0540121843	-0.085186116
## selfemployed_dummy	-0.38427914	-0.347563449	0.1809395855	-0.078377842
## housemaid_dummy	0.06064025	-0.692640876	-0.1893910616	0.035271397
## unemployed_dummy	0.42552982	0.377046144	0.1832868628	-0.110438590
## student_dummy	-0.04496363	0.156838269	-0.0209128774	0.025368516
##	PC17	PC18	PC19	PC20
## age	-0.037160967	-0.011395055	0.004270038	-0.002653752
## balance	0.709471260	-0.015272643	-0.057394328	-0.072379184
## day	-0.009013789	0.128366794	0.624361632	0.106462164

## duration	-0.216033306	-0.232989994	-0.179382182	0.034209294
## campaign	-0.108187382	-0.287716654	-0.633312446	-0.171195787
## pdays	-0.022939603	-0.023669722	0.035965672	-0.081862029
## previous	-0.049672329	-0.252250229	0.027563398	0.414908550
## default_dummy	0.029351020	0.129669889	0.008551487	-0.050989400
## housing_dummy	-0.100569900	-0.197716946	0.244515968	-0.632498855
## loan_dummy	0.543946751	-0.420771568	0.052726883	0.022011974
## cell_dummy	0.119496681	0.505831845	-0.168427786	-0.439011235
## married_dummy	-0.053923228	-0.003352423	0.025539755	-0.028841388
## divorced_dummy	0.053592619	0.070698139	-0.036412103	-0.003909470
## admin_dummy	-0.008312730	0.069075030	-0.079474086	0.050373512
## bluecollar_dummy	0.190613253	0.284032462	-0.124411796	0.129833453
## technician_dummy	-0.023419725	-0.059094194	0.006253981	0.037674849
## services_dummy	0.078145011	0.170717871	-0.092914027	0.089732118
## management_dummy	-0.116261327	-0.159929886	0.076553883	0.014664990
## retired_dummy	-0.100774374	-0.095893822	0.117041096	-0.302847161
## entrepreneur_dummy	-0.172236441	0.094384520	-0.017606901	0.040395179
## selfemployed_dummy	-0.060090151	-0.058492199	0.051416395	-0.017525563
## housemaid_dummy	0.039720413	-0.102089975	0.076401366	-0.148963882
## unemployed_dummy	0.061637150	-0.174740024	0.061557886	-0.066020916
## student_dummy	0.039035968	-0.286235907	0.126297701	-0.167598278
##	PC21	PC22	PC23	PC24
## age	-0.451782216	-0.004216847	0.588238871	-4.830154e-03
## balance	0.043605841	0.005528825	-0.042296191	-4.965605e-05
## day	-0.015452288	-0.056296283	0.009143914	1.183824e-03
## duration	-0.057922928	-0.033580808	-0.003434896	3.072494e-04
## campaign	0.024397810	-0.043478170	-0.003245432	-1.064625e-03
## pdays	-0.044950180	-0.723831081	-0.054510085	-4.102001e-04
## previous	0.084281044	0.564649925	0.006781896	3.270630e-04
## default_dummy	-0.027986450	-0.002612256	0.003942272	3.859072e-04
## housing_dummy	-0.065209293	0.232041590	0.127320981	6.805888e-03
## loan_dummy	0.020819921	-0.047439392	0.031579455	2.413718e-03
## cell_dummy	-0.063014058	0.276869517	0.068198534	3.527206e-03
## married_dummy	-0.406121507	0.068971605	-0.562689452	5.148771e-04
## divorced_dummy	-0.362858870	0.078914538	-0.499066462	2.035925e-03
## admin_dummy	-0.068071222	0.004166520	0.013590402	-3.464943e-01
## bluecollar_dummy	0.017171577	0.020748067	0.052998977	-4.470567e-01
## technician_dummy	-0.087987251	-0.021730367	0.007402561	-4.065986e-01
## services_dummy	-0.051923971	0.005764780	0.039950310	-3.147110e-01
## management_dummy	-0.041621190	-0.037921994	0.003559673	-4.416887e-01
## retired_dummy	0.515997779	0.045103915	-0.236810805	-2.341232e-01
## entrepreneur_dummy	0.062639086	-0.016504655	0.016138950	-1.940375e-01
## selfemployed_dummy	-0.016697782	-0.006007378	0.008764985	-1.995174e-01
## housemaid_dummy	0.129384326	0.007279696	-0.008482480	-1.766470e-01
## unemployed_dummy	-0.008035671	0.005774988	0.003961449	-1.816088e-01
## student_dummy	-0.412647180	0.049405350	0.018443658	-1.551608e-01

summary(bank_pca)

```
## Importance of components:
##          PC1      PC2      PC3      PC4      PC5      PC6
## Standard deviation  1.33845 1.2850 1.24822 1.17257 1.11445 1.09587
## Proportion of Variance 0.07464 0.0688 0.06492 0.05729 0.05175 0.05004
## Cumulative Proportion 0.07464 0.1434 0.20837 0.26566 0.31741 0.36744
##          PC7      PC8      PC9      PC10     PC11     PC12
## Standard deviation  1.0733 1.05902 1.04666 1.04301 1.02140 1.01736
## Proportion of Variance 0.0480 0.04673 0.04565 0.04533 0.04347 0.04313
## Cumulative Proportion 0.4154 0.46217 0.50782 0.55314 0.59661 0.63974
##          PC13     PC14     PC15     PC16     PC17     PC18
## Standard deviation  1.01417 1.0123 0.98721 0.95731 0.94523 0.91926
## Proportion of Variance 0.04286 0.0427 0.04061 0.03818 0.03723 0.03521
## Cumulative Proportion 0.68259 0.7253 0.76590 0.80408 0.84131 0.87652
##          PC19     PC20     PC21     PC22     PC23     PC24
## Standard deviation  0.9086 0.83749 0.76847 0.7099 0.57852 0.08595
## Proportion of Variance 0.0344 0.02922 0.02461 0.0210 0.01395 0.00031
## Cumulative Proportion 0.9109 0.94014 0.96475 0.9858 0.99969 1.00000
```

```
plot(bank_pca)
(eigen_bank <- bank_pca$sdev^2)
```

```
## [1] 1.791447466 1.651313604 1.558058286 1.374921481 1.242001025
## [6] 1.200936637 1.151888232 1.121516977 1.095497180 1.087860709
## [11] 1.043259127 1.035017703 1.028535131 1.024738242 0.974581823
## [16] 0.916439276 0.893461825 0.845033004 0.825488681 0.701397824
## [21] 0.590552567 0.503980892 0.334684406 0.007387903
```

```
names(eigen_bank) <- paste("PC",1:24,sep="")
eigen_bank
```

```
##          PC1      PC2      PC3      PC4      PC5      PC6
## 1.791447466 1.651313604 1.558058286 1.374921481 1.242001025 1.200936637
##          PC7      PC8      PC9      PC10     PC11     PC12
## 1.151888232 1.121516977 1.095497180 1.087860709 1.043259127 1.035017703
##          PC13     PC14     PC15     PC16     PC17     PC18
## 1.028535131 1.024738242 0.974581823 0.916439276 0.893461825 0.845033004
##          PC19     PC20     PC21     PC22     PC23     PC24
## 0.825488681 0.701397824 0.590552567 0.503980892 0.334684406 0.007387903
```

```
head(bank_pca$x)
```

```
##          PC1      PC2      PC3      PC4      PC5      PC6
## [1,] 1.29181970 0.09754913 0.4996106 -0.3536200 -1.6855612 -0.7630621
## [2,] -0.54648140 0.48919637 -0.7014667 0.9386638 0.7378376 1.3554017
## [3,] 0.10724104 1.80356047 0.5513918 0.6028371 0.7512374 -0.7288926
## [4,] 0.70107847 1.95630133 1.3861299 0.7967369 -1.1881234 0.5499114
## [5,] -0.36223921 0.18602263 -0.8376523 0.4469301 -0.9205231 0.9338749
## [6,] -0.02328569 0.68364065 0.2192581 -0.7658717 -1.5875162 -0.7308372
##          PC7      PC8      PC9      PC10     PC11     PC12
## [1,] 0.02445074 -0.69917008 0.7147376 1.18502447 -0.5740131 -0.1948376
## [2,] 0.78526608 -0.63903878 0.6168238 1.30272649 -0.6349214 -0.2779303
```

```
## [3,] -2.21260638 -1.00613898 2.0680727 -1.04227846 0.1106891 -2.8715414
## [4,] 0.73177730 0.06296081 0.2350184 0.16245768 -0.4037046 -0.1789498
## [5,] -0.34264621 -0.13645572 0.2448891 -0.06414585 -0.5349175 -0.1721052
## [6,] -0.16404386 -0.60546901 0.6694297 0.92146462 -0.7797231 -0.1427361
##          PC13          PC14          PC15          PC16          PC17          PC18
## [1,] -0.10201278 -0.4251133 1.2610833 0.35644573 -0.63547357 -1.0866594
## [2,] -0.02102612 -0.5275520 1.6105391 0.32786702 -0.65343429 -0.7195127
## [3,] 3.43844715 -1.3482848 0.8825167 -1.35714665 -0.08868156 -1.1063933
## [4,] -0.01583416 -0.3037606 1.1635790 -0.03536626 0.14597205 0.1655044
## [5,] -0.20697550 -0.2316946 1.0615040 -0.10550810 -0.39585073 -0.1941234
## [6,] -0.21115020 -0.3713310 1.5724255 0.05445035 -0.89816279 -0.9420118
##          PC19          PC20          PC21          PC22          PC23
## [1,] 0.16904841 -0.10576287 -0.986027847 0.03462465 0.6805872
## [2,] 0.05622714 0.05700325 0.300740977 -0.05509552 1.0964283
## [3,] 0.18502281 0.17725006 0.598859767 -0.06384403 -0.5042805
## [4,] -0.19656428 0.16951665 -0.345028865 0.20358858 0.2025300
## [5,] -0.48923760 1.23884612 1.124331478 -0.46578809 0.2107922
## [6,] 0.28082078 -0.07076961 -0.007420435 0.05619476 -0.5653373
##          PC24
## [1,] -0.0149767816
## [2,] -0.0112826192
## [3,] 0.0007625259
## [4,] -0.0119549921
## [5,] 1.0675303226
## [6,] -0.0046291489
```

```
dim(bank_pca$x)
```

```
## [1] 45211    24
```

```
# Identifying the scores by their fixed deposit status
```

```
jobtyp_pca <- data.frame((bank$y), bank_pca$x)
head(jobtyp_pca)
```

```
##   X.bank.y.      PC1      PC2      PC3      PC4      PC5
## 1      no 1.29181970 0.09754913 0.4996106 -0.3536200 -1.6855612
## 2      no -0.54648140 0.48919637 -0.7014667 0.9386638 0.7378376
## 3      no 0.10724104 1.80356047 0.5513918 0.6028371 0.7512374
## 4      no 0.70107847 1.95630133 1.3861299 0.7967369 -1.1881234
## 5      no -0.36223921 0.18602263 -0.8376523 0.4469301 -0.9205231
## 6      no -0.02328569 0.68364065 0.2192581 -0.7658717 -1.5875162
##          PC6      PC7      PC8      PC9      PC10      PC11
## 1 -0.7630621 0.02445074 -0.69917008 0.7147376 1.18502447 -0.5740131
## 2 1.3554017 0.78526608 -0.63903878 0.6168238 1.30272649 -0.6349214
## 3 -0.7288926 -2.21260638 -1.00613898 2.0680727 -1.04227846 0.1106891
## 4 0.5499114 0.73177730 0.06296081 0.2350184 0.16245768 -0.4037046
## 5 0.9338749 -0.34264621 -0.13645572 0.2448891 -0.06414585 -0.5349175
## 6 -0.7308372 -0.16404386 -0.60546901 0.6694297 0.92146462 -0.7797231
##          PC12      PC13      PC14      PC15      PC16      PC17
## 1 -0.1948376 -0.10201278 -0.4251133 1.2610833 0.35644573 -0.63547357
## 2 -0.2779303 -0.02102612 -0.5275520 1.6105391 0.32786702 -0.65343429
```

```
## 3 -2.8715414 3.43844715 -1.3482848 0.8825167 -1.35714665 -0.08868156
## 4 -0.1789498 -0.01583416 -0.3037606 1.1635790 -0.03536626 0.14597205
## 5 -0.1721052 -0.20697550 -0.2316946 1.0615040 -0.10550810 -0.39585073
## 6 -0.1427361 -0.21115020 -0.3713310 1.5724255 0.05445035 -0.89816279
##          PC18          PC19          PC20          PC21          PC22          PC23
## 1 -1.0866594 0.16904841 -0.10576287 -0.986027847 0.03462465 0.6805872
## 2 -0.7195127 0.05622714 0.05700325 0.300740977 -0.05509552 1.0964283
## 3 -1.1063933 0.18502281 0.17725006 0.598859767 -0.06384403 -0.5042805
## 4 0.1655044 -0.19656428 0.16951665 -0.345028865 0.20358858 0.2025300
## 5 -0.1941234 -0.48923760 1.23884612 1.124331478 -0.46578809 0.2107922
## 6 -0.9420118 0.28082078 -0.07076961 -0.007420435 0.05619476 -0.5653373
##          PC24
## 1 -0.0149767816
## 2 -0.0112826192
## 3 0.0007625259
## 4 -0.0119549921
## 5 1.0675303226
## 6 -0.0046291489
```

Means of scores for all the PC's classified by Survival status

```
tabmeansPC <- aggregate(jobtyp_pca[,2:25],by=list(y=bank$y),mean)
tabmeansPC
```

```
##      y          PC1          PC2          PC3          PC4          PC5
## 1 no 0.008469446 0.1080954 -0.01877773 0.004302355 0.03554775
## 2 yes -0.063928385 -0.8159171 0.14173654 -0.032474683 -0.26831865
##          PC6          PC7          PC8          PC9          PC10          PC11
## 1 -0.05612354 0.01339727 0.02935573 -0.03583849 0.001314019 -0.01925154
## 2 0.42362713 -0.10112421 -0.22158050 0.27051317 -0.009918370 0.14531288
##          PC12          PC13          PC14          PC15          PC16          PC17
## 1 -0.003168746 0.02056883 -0.02240821 0.1043351 -0.02350089 0.02628173
## 2 0.023918072 -0.15525594 0.16913983 -0.7875340 0.17738753 -0.19837765
##          PC18          PC19          PC20          PC21          PC22
## 1 0.01537703 0.02477039 -0.01122966 0.000670917 0.007625051
## 2 -0.11606766 -0.18696982 0.08476284 -0.005064162 -0.057554790
##          PC23          PC24
## 1 -0.002090892 0.0001265467
## 2 0.015782304 -0.0009551897
```

```
tabmeansPC <- tabmeansPC[rev(order(tabmeansPC$y)),]
tabmeansPC
```

```
##      y          PC1          PC2          PC3          PC4          PC5
## 2 yes -0.063928385 -0.8159171 0.14173654 -0.032474683 -0.26831865
## 1 no 0.008469446 0.1080954 -0.01877773 0.004302355 0.03554775
##          PC6          PC7          PC8          PC9          PC10          PC11
## 2 0.42362713 -0.10112421 -0.22158050 0.27051317 -0.009918370 0.14531288
## 1 -0.05612354 0.01339727 0.02935573 -0.03583849 0.001314019 -0.01925154
##          PC12          PC13          PC14          PC15          PC16          PC17
## 2 0.023918072 -0.15525594 0.16913983 -0.7875340 0.17738753 -0.19837765
## 1 -0.003168746 0.02056883 -0.02240821 0.1043351 -0.02350089 0.02628173
```



```
##          PC18          PC19          PC20          PC21          PC22
## 2 -0.11606766 -0.18696982  0.08476284 -0.005064162 -0.057554790
## 1  0.01537703  0.02477039 -0.01122966  0.000670917  0.007625051
##          PC23          PC24
## 2  0.015782304 -0.0009551897
## 1 -0.002090892  0.0001265467
```

```
tabfmeans <- t(tabmeansPC[, -1])
tabfmeans
```

```
##          2          1
## PC1 -0.0639283850  0.0084694461
## PC2 -0.8159171247  0.1080954279
## PC3  0.1417365351 -0.0187777299
## PC4 -0.0324746835  0.0043023546
## PC5 -0.2683186490  0.0355477515
## PC6  0.4236271341 -0.0561235387
## PC7 -0.1011242098  0.0133972733
## PC8 -0.2215805050  0.0293557259
## PC9  0.2705131742 -0.0358384895
## PC10 -0.0099183697  0.0013140188
## PC11  0.1453128776 -0.0192515357
## PC12  0.0239180724 -0.0031687462
## PC13 -0.1552559395  0.0205688258
## PC14  0.1691398279 -0.0224082097
## PC15 -0.7875340254  0.1043351400
## PC16  0.1773875282 -0.0235008927
## PC17 -0.1983776471  0.0262817338
## PC18 -0.1160676580  0.0153770313
## PC19 -0.1869698201  0.0247703867
## PC20  0.0847628374 -0.0112296640
## PC21 -0.0050641615  0.0006709170
## PC22 -0.0575547895  0.0076250509
## PC23  0.0157823037 -0.0020908924
## PC24 -0.0009551897  0.0001265467
```

```
colnames(tabfmeans) <- t(as.vector(tabmeansPC[1]))
tabfmeans
```

```
##          yes          no
## PC1 -0.0639283850  0.0084694461
## PC2 -0.8159171247  0.1080954279
## PC3  0.1417365351 -0.0187777299
## PC4 -0.0324746835  0.0043023546
## PC5 -0.2683186490  0.0355477515
## PC6  0.4236271341 -0.0561235387
## PC7 -0.1011242098  0.0133972733
## PC8 -0.2215805050  0.0293557259
## PC9  0.2705131742 -0.0358384895
## PC10 -0.0099183697  0.0013140188
## PC11  0.1453128776 -0.0192515357
```



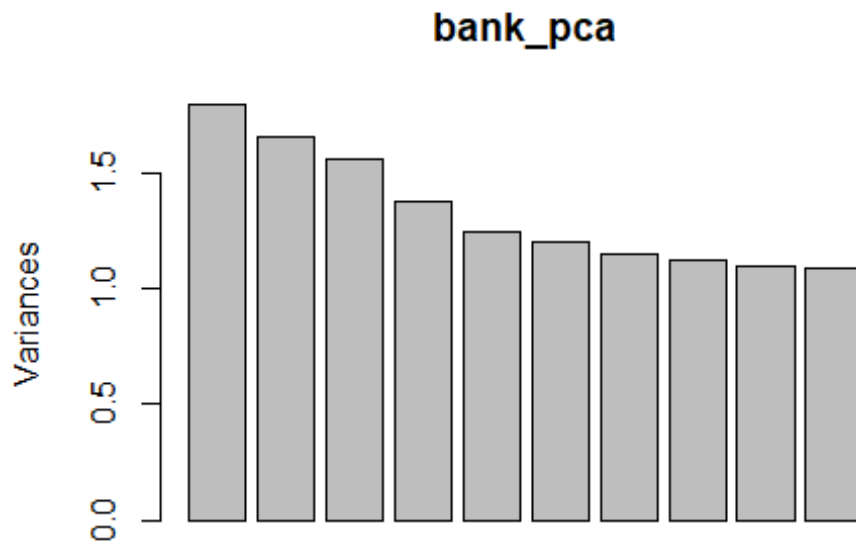
```
## PC12  0.0239180724 -0.0031687462
## PC13 -0.1552559395  0.0205688258
## PC14  0.1691398279 -0.0224082097
## PC15 -0.7875340254  0.1043351400
## PC16  0.1773875282 -0.0235008927
## PC17 -0.1983776471  0.0262817338
## PC18 -0.1160676580  0.0153770313
## PC19 -0.1869698201  0.0247703867
## PC20  0.0847628374 -0.0112296640
## PC21 -0.0050641615  0.0006709170
## PC22 -0.0575547895  0.0076250509
## PC23  0.0157823037 -0.0020908924
## PC24 -0.0009551897  0.0001265467
```

Standard deviations of scores for all the PC's classified by Survival status

```
tabsdsPC <- aggregate(jobtyp_pca[,2:25],by=list(y=bank$y),sd)
tabfsds <- t(tabsdsPC[, -1])
colnames(tabfsds) <- t(as.vector(tabsdsPC[1]))
tabfsds
```

```
##           no           yes
## PC1  1.28476771 1.68826923
## PC2  1.24695524 1.27492988
## PC3  1.23925152 1.30539417
## PC4  1.16170608 1.25117631
## PC5  1.11382274 1.08224584
## PC6  1.07066853 1.18748749
## PC7  1.08430963 0.98007798
## PC8  1.05207177 1.08480619
## PC9  1.04383899 1.02826924
## PC10 1.03278502 1.11719325
## PC11 1.01060916 1.08861504
## PC12 0.99931240 1.14424863
## PC13 1.01802624 0.97068137
## PC14 1.01090791 1.00682254
## PC15 0.88502357 1.31017633
## PC16 0.96805206 0.85135680
## PC17 0.93631498 0.98777708
## PC18 0.91774413 0.92244838
## PC19 0.91092635 0.86810144
## PC20 0.83709260 0.83575060
## PC21 0.75403431 0.86983129
## PC22 0.69194168 0.83099323
## PC23 0.57875666 0.57653320
## PC24 0.08586204 0.08663796
```

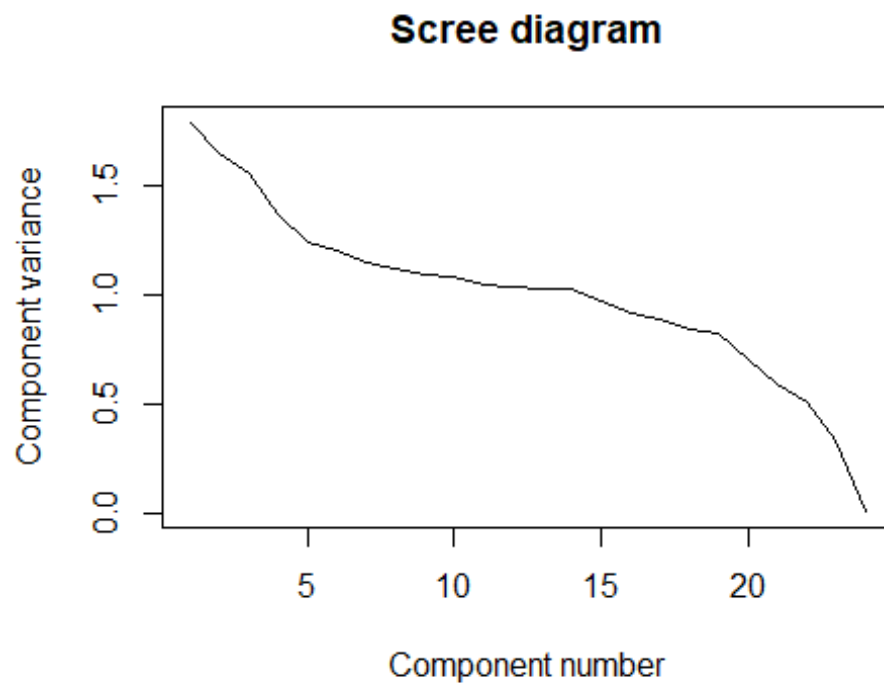
```
plot(bank_pca)
```



```
summary(bank_pca)
```

```
## Importance of components:
##               PC1      PC2      PC3      PC4      PC5      PC6
## Standard deviation  1.33845 1.2850 1.24822 1.17257 1.11445 1.09587
## Proportion of Variance 0.07464 0.0688 0.06492 0.05729 0.05175 0.05004
## Cumulative Proportion 0.07464 0.1434 0.20837 0.26566 0.31741 0.36744
##               PC7      PC8      PC9      PC10     PC11     PC12
## Standard deviation  1.0733 1.05902 1.04666 1.04301 1.02140 1.01736
## Proportion of Variance 0.0480 0.04673 0.04565 0.04533 0.04347 0.04313
## Cumulative Proportion 0.4154 0.46217 0.50782 0.55314 0.59661 0.63974
##               PC13     PC14     PC15     PC16     PC17     PC18
## Standard deviation  1.01417 1.0123 0.98721 0.95731 0.94523 0.91926
## Proportion of Variance 0.04286 0.0427 0.04061 0.03818 0.03723 0.03521
## Cumulative Proportion 0.68259 0.7253 0.76590 0.80408 0.84131 0.87652
##               PC19     PC20     PC21     PC22     PC23     PC24
## Standard deviation  0.9086 0.83749 0.76847 0.7099 0.57852 0.08595
## Proportion of Variance 0.0344 0.02922 0.02461 0.0210 0.01395 0.00031
## Cumulative Proportion 0.9109 0.94014 0.96475 0.9858 0.99969 1.00000
```

```
plot(eigen_bank, xlab = "Component number", ylab = "Component variance", type  
= "l", main = "Scree diagram")
```

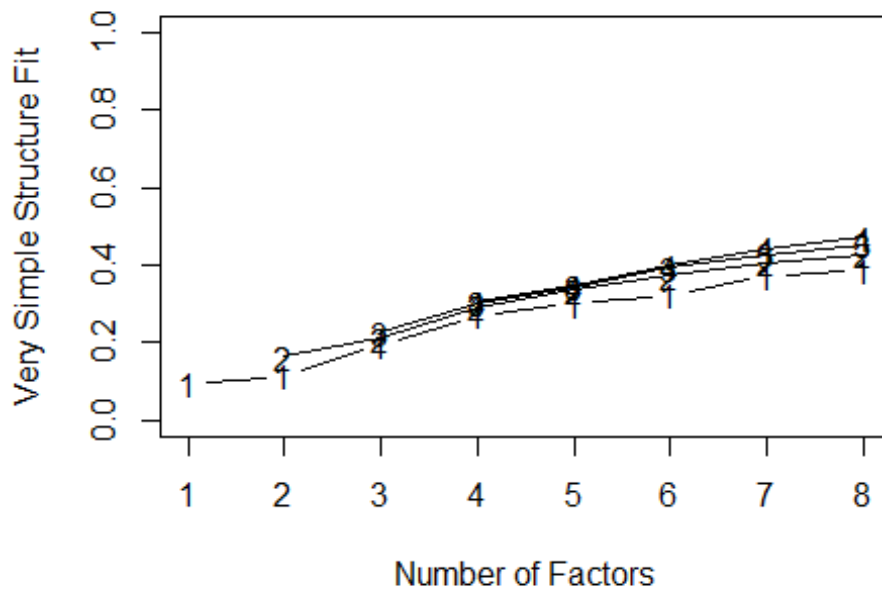


#Can take till PC16 i.e. here 24 features reduced to 16

```
library(psych)
```

```
vss(bank1)
```

Very Simple Structure



```
##
## Very Simple Structure
## Call: vss(x = bank1)
## VSS complexity 1 achieves a maximum of 0.39 with 8 factors
## VSS complexity 2 achieves a maximum of 0.43 with 8 factors
##
## The Velicer MAP achieves a minimum of 0.01 with 1 factors
## BIC achieves a minimum of 127809 with 8 factors
## Sample Size adjusted BIC achieves a minimum of 128164.9 with 8 factors
##
## Statistics by number of factors
##   vss1 vss2   map dof  chisq prob sqresid  fit RMSEA   BIC  SABIC
## 1 0.094 0.00 0.0088 252 222228    0      25 0.094  0.14 219527 220328
## 2 0.115 0.17 0.0111 229 210450    0      23 0.165  0.14 207995 208723
## 3 0.195 0.21 0.0130 207 195959    0      21 0.227  0.14 193740 194398
## 4 0.271 0.29 0.0151 186 181482    0      19 0.305  0.15 179488 180079
## 5 0.301 0.34 0.0189 166 176060    0      18 0.348  0.15 174281 174808
## 6 0.323 0.37 0.0238 147 159723    0      17 0.400  0.15 158147 158614
## 7 0.372 0.41 0.0302 129 148403    0      15 0.449  0.16 147020 147430
## 8 0.390 0.43 0.0372 112 129010    0      14 0.490  0.16 127809 128165
##   complex eChisq  SRMR eCRMS  eBIC
## 1      1.0 125527 0.071 0.074 122826
## 2      1.4 101991 0.064 0.070  99536
## 3      1.6  83571 0.058 0.067  81352
## 4      1.6  55607 0.047 0.058  53613
## 5      1.7  49688 0.045 0.058  47909
## 6      1.7  33856 0.037 0.050  32280
```

```
## 7      1.7  27018 0.033 0.048  25635
## 8      1.9  18021 0.027 0.042  16820
```

#therefore choosing 8 Factors

#Oblique rotation

```
fit.pc <- fa(bank1, nfactors=8, rotate="oblimin")
```

```
fit.pc
```

```
## Factor Analysis using method = minres
```

```
## Call: fa(r = bank1, nfactors = 8, rotate = "oblimin")
```

```
## Standardized loadings (pattern matrix) based upon correlation matrix
```

```
##           MR1    MR7    MR2    MR4    MR3    MR8    MR5    MR6    h2
## age           0.99  0.03  0.00  0.00  0.00  0.00  0.00 -0.01 0.9993
## balance        0.07  0.02 -0.10  0.02  0.01 -0.06 -0.06 -0.07 0.0224
## day           -0.01  0.02 -0.01 -0.11  0.03  0.04  0.00  0.00 0.0150
## duration        0.00 -0.03 -0.01  0.02 -0.03 -0.03 -0.03 -0.02 0.0024
## campaign       -0.01  0.04  0.03 -0.11  0.05  0.05  0.01  0.02 0.0160
## pdays          0.00  0.00  0.01  0.95  0.00  0.00  0.00  0.00 0.8968
## previous        0.01  0.01 -0.04  0.48  0.01 -0.01 -0.01 -0.02 0.2344
## default_dummy  -0.01 -0.02  0.03 -0.04  0.01  0.01  0.01  0.02 0.0027
## housing_dummy  -0.12  0.02  0.32  0.08  0.10  0.13  0.16  0.18 0.1361
## loan_dummy      0.01  0.03  0.08 -0.03  0.03  0.07  0.08  0.08 0.0133
## cell_dummy     -0.07  0.00 -0.19  0.24  0.02 -0.01 -0.06 -0.09 0.0984
## married_dummy   0.05  0.99  0.02  0.00  0.01  0.01  0.01  0.01 1.0005
## divorced_dummy  0.30 -0.52  0.08 -0.01  0.06  0.06  0.07  0.08 0.2901
## admin_dummy    -0.01  0.00 -0.03  0.00 -0.02 -0.02  0.99 -0.02 1.0042
## bluecollar_dummy -0.01  0.00  0.97  0.00 -0.04 -0.04 -0.04 -0.04 1.0001
## technician_dummy -0.01  0.00 -0.03  0.00 -0.03  0.99 -0.02 -0.02 1.0150
## services_dummy -0.01  0.00 -0.03  0.00 -0.02 -0.02 -0.02  0.99 0.9985
## management_dummy -0.01  0.00 -0.03  0.00  0.99 -0.03 -0.02 -0.02 1.0250
## retired_dummy   0.42 -0.02 -0.17  0.02 -0.17 -0.14 -0.12 -0.10 0.2444
## entrepreneur_dummy -0.02  0.04 -0.09 -0.01 -0.09 -0.08 -0.06 -0.05 0.0136
## selfemployed_dummy -0.05  0.03 -0.13 -0.01 -0.13 -0.12 -0.09 -0.09 0.0258
## housemaid_dummy  0.04  0.02 -0.11 -0.03 -0.10 -0.09 -0.07 -0.07 0.0234
## unemployed_dummy -0.03 -0.01 -0.12 -0.01 -0.12 -0.11 -0.09 -0.08 0.0233
## student_dummy  -0.23 -0.09 -0.21  0.03 -0.20 -0.19 -0.17 -0.15 0.1155
```

```
##           u2 com
## age           0.00071 1.0
## balance        0.97764 4.6
## day           0.98499 1.6
## duration        0.99758 5.3
## campaign       0.98402 2.4
## pdays          0.10319 1.0
## previous        0.76563 1.0
## default_dummy   0.99727 3.4
## housing_dummy   0.86390 3.4
## loan_dummy      0.98667 4.6
## cell_dummy      0.90163 2.6
```

```

## married_dummy      -0.00051 1.0
## divorced_dummy     0.70992 1.8
## admin_dummy        -0.00420 1.0
## bluecollar_dummy   -0.00012 1.0
## technician_dummy   -0.01503 1.0
## services_dummy     0.00147 1.0
## management_dummy   -0.02504 1.0
## retired_dummy      0.75563 2.4
## entrepreneur_dummy 0.98637 4.9
## selfemployed_dummy 0.97419 5.0
## housemaid_dummy    0.97659 5.2
## unemployed_dummy   0.97672 4.8
## student_dummy      0.88454 6.1
##
##
##          MR1  MR7  MR2  MR4  MR3  MR8  MR5  MR6
## SS loadings      1.33 1.24 1.17 1.22 1.08 1.07 1.06 1.06
## Proportion Var    0.06 0.05 0.05 0.05 0.04 0.04 0.04 0.04
## Cumulative Var    0.06 0.11 0.16 0.21 0.25 0.30 0.34 0.38
## Proportion Explained 0.14 0.14 0.13 0.13 0.12 0.12 0.11 0.11
## Cumulative Proportion 0.14 0.28 0.41 0.54 0.65 0.77 0.89 1.00
##
## With factor correlations of
##          MR1  MR7  MR2  MR4  MR3  MR8  MR5  MR6
## MR1  1.00  0.21 -0.06 -0.03 -0.03 -0.07 -0.06 -0.06
## MR7  0.21  1.00  0.12 -0.03 -0.02 -0.05 -0.05 -0.01
## MR2 -0.06  0.12  1.00  0.02 -0.21 -0.19 -0.15 -0.13
## MR4 -0.03 -0.03  0.02  1.00  0.00 -0.01  0.03  0.01
## MR3 -0.03 -0.02 -0.21  0.00  1.00 -0.18 -0.14 -0.13
## MR8 -0.07 -0.05 -0.19 -0.01 -0.18  1.00 -0.12 -0.11
## MR5 -0.06 -0.05 -0.15  0.03 -0.14 -0.12  1.00 -0.09
## MR6 -0.06 -0.01 -0.13  0.01 -0.13 -0.11 -0.09  1.00
##
## Mean item complexity = 2.8
## Test of the hypothesis that 8 factors are sufficient.
##
## The degrees of freedom for the null model are 276 and the objective
function was 5.35 with Chi Square of 241726.1
## The degrees of freedom for the model are 112 and the objective function
was 2.85
##
## The root mean square of the residuals (RMSR) is 0.03
## The df corrected root mean square of the residuals is 0.04
##
## The harmonic number of observations is 45211 with the empirical chi
square 18021.02 with prob < 0
## The total number of observations was 45211 with Likelihood Chi Square =
129009.6 with prob < 0
##
## Tucker Lewis Index of factoring reliability = -0.316
## RMSEA index = 0.16 and the 90 % confidence intervals are 0.159 0.16

```

```
## BIC = 127809
## Fit based upon off diagonal values = 0.89

#what are factors for each variable , what is affectin and
#Fit based upon off diagonal values = 0.89 higher the better
#residual should be less
round(fit.pc$values, 3)

## [1] 1.577 1.398 1.257 1.206 1.165 1.109 1.011 0.562 0.261 0.213
## [11] 0.138 0.122 0.080 0.054 0.042 0.005 -0.003 -0.033 -0.048 -0.065
## [21] -0.097 -0.147 -0.183 -0.408

fit.pc$loadings

##
## Loadings:
##
##          MR1    MR7    MR2    MR4    MR3    MR8    MR5    MR6
## age          0.992
## balance
## day              -0.108
## duration
## campaign        -0.108
## pdays          0.947
## previous        0.484
## default_dummy
## housing_dummy   -0.121          0.318          0.127 0.163 0.181
## loan_dummy
## cell_dummy      -0.193 0.236
## married_dummy          0.987
## divorced_dummy   0.301 -0.516
## admin_dummy
## bluecollar_dummy          0.971
## technician_dummy          0.990
## services_dummy          0.987
## management_dummy          0.994
## retired_dummy    0.416          -0.169          -0.173 -0.143 -0.121 -0.103
## entrepreneur_dummy
## selfemployed_dummy          -0.133          -0.127 -0.115
## housemaid_dummy          -0.109          -0.103
## unemployed_dummy          -0.124          -0.120 -0.111
## student_dummy    -0.227          -0.206          -0.197 -0.193 -0.167 -0.153
##
##          MR1    MR7    MR2    MR4    MR3    MR8    MR5    MR6
## SS loadings  1.333 1.257 1.235 1.222 1.128 1.116 1.096 1.093
## Proportion Var 0.056 0.052 0.051 0.051 0.047 0.046 0.046 0.046
## Cumulative Var 0.056 0.108 0.159 0.210 0.257 0.304 0.349 0.395

#View(fit.pc)
# Loadings with more digits
for (i in c(1,2,3,4,5,6,7,8)) { print(fit.pc$loadings[[1,i]])}
```

```
## [1] 0.9919925
## [1] 0.03003172
## [1] -0.002983055
## [1] 0.002011022
## [1] 0.001468767
## [1] -0.002941017
## [1] -0.004388868
## [1] -0.006695473
```

Communalities

```
fit.pc$communality
```

##	age	balance	day
##	0.999288205	0.022361350	0.015006623
##	duration	campaign	pdays
##	0.002422461	0.015976188	0.896806997
##	previous	default_dummy	housing_dummy
##	0.234369348	0.002727426	0.136103920
##	loan_dummy	cell_dummy	married_dummy
##	0.013326636	0.098368395	1.000510588
##	divorced_dummy	admin_dummy	bluecollar_dummy
##	0.290078947	1.004202170	1.000119833
##	technician_dummy	services_dummy	management_dummy
##	1.015031575	0.998527223	1.025044500
##	retired_dummy	entrepreneur_dummy	selfemployed_dummy
##	0.244367140	0.013634905	0.025814261
##	housemaid_dummy	unemployed_dummy	student_dummy
##	0.023408655	0.023284051	0.115458229

Rotated factor scores, Notice the columns ordering: MR1, MR7, MR2, MR4, MR3m

```
fit.pc$scores
```

##		MR1	MR7	MR2	MR4	MR3
##	[1,]	1.566961e+00	0.7271740	-2.756735e-01	-0.389091132	2.074649e+00
##	[2,]	3.231035e-01	-1.2902486	-3.272992e-01	-0.387805588	-3.809572e-01
##	[3,]	-8.012463e-01	0.7951218	-1.187477e-01	-0.418728529	-3.028731e-01
##	[4,]	5.194126e-01	0.7787244	2.024276e+00	-0.379349316	-3.675033e-01
##	[5,]	1.664591e+00	-1.5823885	-1.295277e+01	-1.636540506	-1.501140e+01
##	[6,]	-6.133475e-01	0.8526226	-3.240004e-01	-0.386638504	1.978772e+00
##	[7,]	-1.161345e+00	-1.2316997	-4.095467e-01	-0.381816152	1.870837e+00
##		MR8	MR5	MR6		
##	[1,]	-2.821297e-01	-1.948512e-01	-0.1662269651		
##	[2,]	2.280185e+00	-2.460201e-01	-0.2098633191		
##	[3,]	-2.050191e-01	-9.325430e-02	-0.0457333141		
##	[4,]	-3.276992e-01	-2.473323e-01	-0.2176696218		
##	[5,]	-1.286523e+01	-1.012301e+01	-8.6933086022		
##	[6,]	-3.522338e-01	-2.443516e-01	-0.2032879385		
##	[7,]	-4.408910e-01	-3.167783e-01	-0.2616576853		
##	[8,]	-1.702203e-01	-4.475210e-02	0.0056285943		

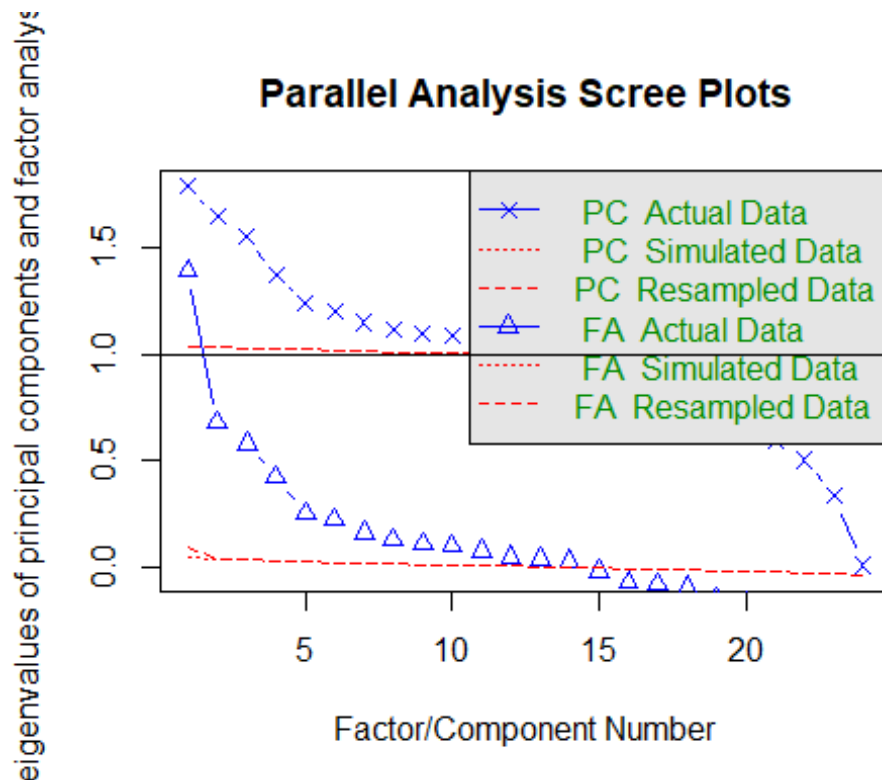

```
##      [9,] -3.588582e-01 -2.514857e-01 -0.2063120899
```

```
fa.parallel(bank1) # See factor recommendation
```

```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate =  
## rotate, : A loading greater than abs(1) was detected. Examine the loadings  
## carefully.
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs  
## = np.obs, : The estimated weights for the factor scores are probably  
## incorrect. Try a different factor extraction method.
```

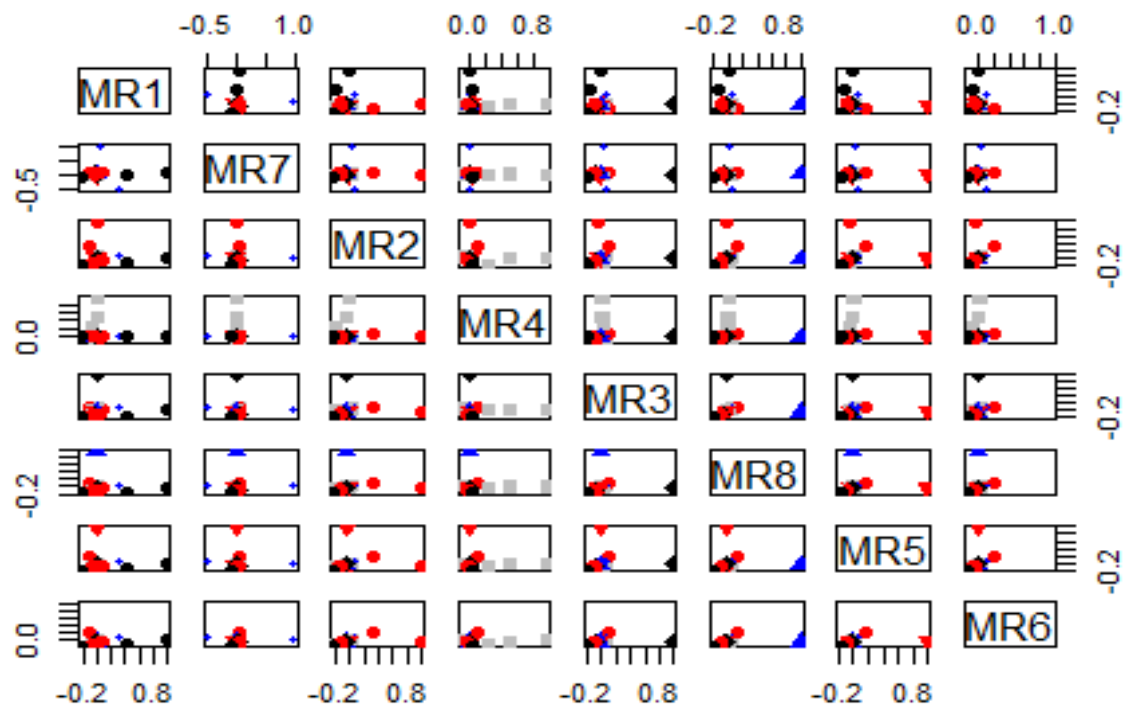
```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate =  
## rotate, : An ultra-Heywood case was detected. Examine the results  
carefully
```



```
## Parallel analysis suggests that the number of factors = 14 and the  
number of components = 14
```

```
fa.plot(fit.pc) # See Correlations within Factors
```

Factor Analysis



```
fa.diagram(fit.pc) # Visualize the relationship
```

Plot Zoom

Factor Analysis

