

# INDIAN CREDIT RISK ANALYSIS

## BUSINESS REPORT

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# PROBLEM STATEMENT

Data description - Please watch the video- Default Risk Estimation-Part-1. After removing variables for multicollinearity, we should try to take at least one variable for creating the model from each of the 4 factors namely -

- 1) Profitability
- 2) Leverage
- 3) Liquidity
- 4) Company's size

In Dr Sarkar's video of Default Risk Estimation, he has clearly bifurcated all the variables in different buckets.

Creation of new variables - This is an important step in the project as the company which is the biggest in size, will also have bigger asset size, cash flows etc. (Hint: We need to think in terms of ratios - Equity to asset ratio, debt to equity ratio etc)

Dependant variable - We need to create a default variable which should take the value of 1 when net worth is negative & 0 when net worth is positive.

# Indian Credit Risk Analysis

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```
library(glmnet)
library(bnstruct)
library(mice)
library(car)
library(tidyverse)
library(corrplot)
library(MASS)
library(psych)
library(GPArotation)
library(car)
library(psy)
library(caret)

raw_data = read.csv('raw-data.csv')

dim(raw_data)
## [1] 3541 52

names(raw_data)
## [1] "Num"
## [2] "Networth.Next.Year"
## [3] "Total.assets"
## [4] "Net.worth"
## [5] "Total.income"
## [6] "Change.in.stock"
## [7] "Total.expenses"
## [8] "Profit.after.tax"
## [9] "PBDITA"
## [10] "PBT"
## [11] "Cash.profit"
## [12] "PBDITA.as...of.total.income"
## [13] "PBT.as...of.total.income"
## [14] "PAT.as...of.total.income"
```

```
## [15] "Cash.profit.as...of.total.income"
## [16] "PAT.as...of.net.worth"
## [17] "Sales"
## [18] "Income.from.financial.services"
## [19] "Other.income"
## [20] "Total.capital"
## [21] "Reserves.and.funds"
## [22] "Deposits..accepted.by.commercial.banks."
## [23] "Borrowings"
## [24] "Current.liabilities...provisions"
## [25] "Deferred.tax.liability"
## [26] "Shareholders.funds"
## [27] "Cumulative.retained.profits"
## [28] "Capital.employed"
## [29] "TOL.TNW"
## [30] "Total.term.liabilities...tangible.net.worth"
## [31] "Contingent.liabilities...Net.worth...."
## [32] "Contingent.liabilities"
## [33] "Net.fixed.assets"
## [34] "Investments"
## [35] "Current.assets"
## [36] "Net.working.capital"
## [37] "Quick.ratio..times."
## [38] "Current.ratio..times."
## [39] "Debt.to.equity.ratio..times."
## [40] "Cash.to.current.liabilities..times."
## [41] "Cash.to.average.cost.of.sales.per.day"
## [42] "Creditors.turnover"
## [43] "Debtors.turnover"
## [44] "Finished.goods.turnover"
## [45] "WIP.turnover"
## [46] "Raw.material.turnover"
## [47] "Shares.outstanding"
## [48] "Equity.face.value"
## [49] "EPS"
## [50] "Adjusted.EPS"
## [51] "Total.liabilities"
## [52] "PE.on.BSE"
```

```
summary(raw_data)
```

```
##      Num      Networth.Next.Year  Total.assets      Net.worth
## Min.   :    1  Min.   :-74265.6  Min.   :    0.1  Min.   :    0.0
## 1st Qu.: 886  1st Qu.:   31.7  1st Qu.:   91.3  1st Qu.:   31.3
## Median :1773  Median :   116.3  Median :   309.7  Median :   102.3
## Mean   :1772  Mean   :  1616.3  Mean   :  3443.4  Mean   :  1295.9
## 3rd Qu.:2658  3rd Qu.:   456.1  3rd Qu.:  1098.7  3rd Qu.:   377.3
## Max.   :3545  Max.   :805773.4  Max.   :1176509.2  Max.   :613151.6
##
##      Total.income      Change.in.stock      Total.expenses
## Min.   :    0.0  Min.   :-3029.40  Min.   :   -0.1
## 1st Qu.:  106.5  1st Qu.:   -1.80  1st Qu.:   95.8
## Median :   444.9  Median :    1.60  Median :   407.7
## Mean   :  4582.8  Mean   :   41.49  Mean   :  4262.9
## 3rd Qu.: 1440.9  3rd Qu.:   18.05  3rd Qu.:  1359.8
```

## Max. :2442828.2	Max. :14185.50	Max. :2366035.3
## NA's :198	NA's :458	NA's :139
## Profit.after.tax	PBDITA	PBT
## Min. : -3908.30	Min. : -440.7	Min. : -3894.80
## 1st Qu.: 0.50	1st Qu.: 6.9	1st Qu.: 0.70
## Median : 8.80	Median : 35.4	Median : 12.40
## Mean : 277.36	Mean : 578.1	Mean : 383.81
## 3rd Qu.: 52.27	3rd Qu.: 150.2	3rd Qu.: 71.97
## Max. :119439.10	Max. :208576.5	Max. :145292.60
## NA's :131	NA's :131	NA's :131
## Cash.profit	PBDITA.as...of.total.income	PBT.as...of.total.income
## Min. : -2245.70	Min. : -6400.000	Min. : -21340.00
## 1st Qu.: 2.90	1st Qu.: 5.000	1st Qu.: 0.55
## Median : 18.85	Median : 9.660	Median : 3.31
## Mean : 392.07	Mean : 4.571	Mean : -17.28
## 3rd Qu.: 93.20	3rd Qu.: 16.390	3rd Qu.: 8.80
## Max. :176911.80	Max. : 100.000	Max. : 100.00
## NA's :131	NA's :68	NA's :68
## PAT.as...of.total.income	Cash.profit.as...of.total.income	
## Min. : -21340.00	Min. : -15020.000	
## 1st Qu.: 0.35	1st Qu.: 2.020	
## Median : 2.34	Median : 5.640	
## Mean : -19.20	Mean : -8.229	
## 3rd Qu.: 6.34	3rd Qu.: 10.700	
## Max. : 150.00	Max. : 100.000	
## NA's :68	NA's :68	
## PAT.as...of.net.worth	Sales	Income.from.financial.services
## Min. : -748.72	Min. : 0.1	Min. : 0.00
## 1st Qu.: 0.00	1st Qu.: 112.7	1st Qu.: 0.40
## Median : 7.92	Median : 453.1	Median : 1.80
## Mean : 10.27	Mean : 4549.5	Mean : 80.84
## 3rd Qu.: 20.19	3rd Qu.: 1433.5	3rd Qu.: 9.68
## Max. :2466.67	Max. :2384984.4	Max. :51938.20
##	NA's :259	NA's :935
## Other.income	Total.capital	Reserves.and.funds
## Min. : 0.00	Min. : 0.1	Min. : -6525.9
## 1st Qu.: 0.40	1st Qu.: 13.1	1st Qu.: 5.0
## Median : 1.40	Median : 42.1	Median : 54.8
## Mean : 41.36	Mean : 216.6	Mean : 1163.8
## 3rd Qu.: 5.97	3rd Qu.: 100.3	3rd Qu.: 277.3
## Max. :42856.70	Max. :78273.2	Max. :625137.8
## NA's :1295	NA's :4	NA's :85
## Deposits..accepted.by.commercial.banks.	Borrowings	
## Mode:logical	Min. : 0.10	
## NA's:3541	1st Qu.: 23.95	
##	Median : 99.20	
##	Mean : 1122.28	
##	3rd Qu.: 352.60	
##	Max. :278257.30	
##	NA's :366	
## Current.liabilities...provisions	Deferred.tax.liability	
## Min. : 0.1	Min. : 0.1	
## 1st Qu.: 17.8	1st Qu.: 3.2	
## Median : 69.4	Median : 13.4	
## Mean : 940.6	Mean : 227.2	

```

## 3rd Qu.: 261.7          3rd Qu.: 50.0
## Max. :352240.3        Max. :72796.6
## NA's :96              NA's :1140
## Shareholders.funds Cumulative.retained.profits Capital.employed
## Min. : 0.0 Min. : -6534.3 Min. : 0.0
## 1st Qu.: 32.0 1st Qu.: 1.1 1st Qu.: 60.8
## Median : 105.6 Median : 37.1 Median : 214.7
## Mean : 1322.1 Mean : 890.5 Mean : 2328.3
## 3rd Qu.: 393.2 3rd Qu.: 202.3 3rd Qu.: 767.3
## Max. :613151.6 Max. :390133.8 Max. :891408.9
## NA's :38
## TOL.TNW Total.term.liabilities...tangible.net.worth
## Min. : -350.480 Min. : -325.600
## 1st Qu.: 0.600 1st Qu.: 0.050
## Median : 1.430 Median : 0.340
## Mean : 3.994 Mean : 1.844
## 3rd Qu.: 2.830 3rd Qu.: 1.000
## Max. : 473.000 Max. : 456.000
##
## Contingent.liabilities...Net.worth.... Contingent.liabilities
## Min. : 0.00 Min. : 0.1
## 1st Qu.: 0.00 1st Qu.: 6.3
## Median : 5.33 Median : 38.0
## Mean : 53.94 Mean : 932.9
## 3rd Qu.: 30.76 3rd Qu.: 192.7
## Max. :14704.27 Max. :559506.8
## NA's :1188
## Net.fixed.assets Investments Current.assets
## Min. : 0.0 Min. : 0.00 Min. : 0.1
## 1st Qu.: 26.0 1st Qu.: 1.00 1st Qu.: 36.2
## Median : 93.5 Median : 8.35 Median : 145.1
## Mean : 1189.7 Mean : 694.73 Mean : 1293.4
## 3rd Qu.: 344.9 3rd Qu.: 64.30 3rd Qu.: 502.2
## Max. :636604.6 Max. :199978.60 Max. :354815.2
## NA's :118 NA's :1435 NA's :66
## Net.working.capital Quick.ratio..times. Current.ratio..times.
## Min. : -63839.0 Min. : 0.000 Min. : 0.00
## 1st Qu.: -1.1 1st Qu.: 0.410 1st Qu.: 0.93
## Median : 16.2 Median : 0.670 Median : 1.23
## Mean : 138.6 Mean : 1.401 Mean : 2.13
## 3rd Qu.: 84.2 3rd Qu.: 1.030 3rd Qu.: 1.71
## Max. : 85782.8 Max. :341.000 Max. :505.00
## NA's :32 NA's :93 NA's :93
## Debt.to.equity.ratio..times. Cash.to.current.liabilities..times.
## Min. : 0.00 Min. : 0.0000
## 1st Qu.: 0.22 1st Qu.: 0.0200
## Median : 0.79 Median : 0.0700
## Mean : 2.78 Mean : 0.4904
## 3rd Qu.: 1.75 3rd Qu.: 0.1900
## Max. :456.00 Max. :165.0000
## NA's :93
## Cash.to.average.cost.of.sales.per.day Creditors.turnover
## Min. : 0.00 Min. : 0.000
## 1st Qu.: 2.79 1st Qu.: 3.700
## Median : 8.03 Median : 6.095

```

```
## Mean      : 158.44                      Mean      : 15.446
## 3rd Qu.:   21.79                      3rd Qu.:  11.490
## Max.     :128040.76                    Max.     :2401.000
## NA's     :85                          NA's     :333
## Debtors.turnover Finished.goods.turnover WIP.turnover
## Min.      : 0.00      Min.      : -0.09      Min.      : -0.18
## 1st Qu.:   3.76      1st Qu.:   8.20      1st Qu.:   5.10
## Median    :  6.32      Median    : 17.27      Median    :  9.76
## Mean      : 17.04      Mean      : 87.08      Mean      : 27.93
## 3rd Qu.:  11.68      3rd Qu.:  40.35      3rd Qu.:  20.24
## Max.      :3135.20     Max.      :17947.60     Max.      :5651.40
## NA's      :328        NA's      :740        NA's      :640
## Raw.material.turnover Shares.outstanding Equity.face.value
## Min.      : -2.00      Min.      :-2.147e+09    Min.      : -999999
## 1st Qu.:    2.99      1st Qu.: 1.316e+06      1st Qu.:    10
## Median    :  6.40      Median    : 4.672e+06    Median    :    10
## Mean      : 19.09      Mean      : 2.207e+07    Mean      : -1334
## 3rd Qu.:  11.85      3rd Qu.: 1.065e+07      3rd Qu.:    10
## Max.      :21092.00     Max.      : 4.130e+09    Max.      :100000
## NA's      :361        NA's      :692        NA's      :692
## EPS Adjusted.EPS Total.liabilities
## Min.      :-843181.8    Min.      :-843181.8    Min.      :  0.1
## 1st Qu.:    0.0        1st Qu.:    0.0        1st Qu.:   91.3
## Median    :  1.4        Median    :  1.2        Median    : 309.7
## Mean      : -220.3      Mean      : -221.5      Mean      : 3443.4
## 3rd Qu.:    9.6        3rd Qu.:    7.5        3rd Qu.: 1098.7
## Max.      : 34522.5     Max.      : 34522.5     Max.      :1176509.2
##
## PE.on.BSE
## Min.      :-1116.64
## 1st Qu.:    3.27
## Median    :  9.10
## Mean      :  63.91
## 3rd Qu.:   17.79
## Max.      :51002.74
## NA's      :2194
```

```
head(raw_data)
```

```
## Num Networth.Next.Year Total.assets Net.worth Total.income
## 1 1 8890.6 17512.3 7093.2 24965.2
## 2 2 394.3 941.0 351.5 1527.4
## 3 3 92.2 232.8 100.6 477.3
## 4 4 2.7 2.7 2.7 NA
## 5 5 109.0 478.5 107.6 1580.5
## 6 6 688.6 2434.4 675.8 2648.6
## Change.in.stock Total.expenses Profit.after.tax PBDITA PBT
## 1 235.8 23657.8 1543.2 2860.2 2417.2
## 2 42.7 1454.9 115.2 283.0 188.4
## 3 -5.2 478.7 -6.6 5.8 -6.6
## 4 NA NA NA NA NA
## 5 -17.0 1558.0 5.5 31.0 6.3
## 6 62.3 2636.4 74.5 200.1 74.5
## Cash.profit PBDITA.as...of.total.income PBT.as...of.total.income
## 1 1872.8 11.46 9.68
```

## 2	158.6		18.53		12.33
## 3	0.3		1.22		-1.38
## 4	NA		0.00		0.00
## 5	11.9		1.96		0.40
## 6	146.9		7.55		2.81
##	PAT.as...of.total.income		Cash.profit.as...of.total.income		
## 1		6.18		7.50	
## 2		7.54		10.38	
## 3		-1.38		0.06	
## 4		0.00		0.00	
## 5		0.35		0.75	
## 6		2.81		5.55	
##	PAT.as...of.net.worth		Sales		Income.from.financial.services
## 1		23.78	24458.0		158.0
## 2		38.08	1504.3		4.0
## 3		-6.35	475.6		1.5
## 4		0.00	NA		NA
## 5		5.25	1575.1		3.9
## 6		21.78	2639.5		6.4
##	Other.income		Total.capital		Reserves.and.funds
## 1		297.2	423.8		6822.8
## 2		15.9	115.5		257.8
## 3		0.2	81.4		19.2
## 4		NA	0.5		2.2
## 5		0.9	6.2		161.8
## 6		0.2	33.8		972.0
##	Deposits..accepted.by.commercial.banks.		Borrowings		
## 1			NA		14.9
## 2			NA		272.5
## 3			NA		35.4
## 4			NA		NA
## 5			NA		193.1
## 6			NA		717.1
##	Current.liabilities...provisions		Deferred.tax.liability		
## 1			9965.9		284.9
## 2			210.0		85.2
## 3			96.8		NA
## 4			NA		NA
## 5			112.8		4.6
## 6			555.9		54.4
##	Shareholders.funds		Cumulative.retained.profits		Capital.employed
## 1		7093.2		6263.3	7108.1
## 2		351.5		247.4	624.0
## 3		100.6		32.4	136.0
## 4		2.7		2.2	2.7
## 5		107.6		82.7	300.7
## 6		698.2		317.7	1415.3
##	Total.term.liabilities...tangible.net.worth				TOL.TNW
## 1				0.00	1.33
## 2				0.34	1.23
## 3				0.29	1.44
## 4				0.00	0.00
## 5				1.59	2.83
## 6				0.37	1.80
##	Contingent.liabilities...Net.worth....		Contingent.liabilities		



## 1			14.80	1049.7
## 2			19.23	67.6
## 3			45.83	46.1
## 4			0.00	NA
## 5			34.94	37.6
## 6			36.28	245.2
##	Net.fixed.assets	Investments	Current.assets	Net.working.capital
## 1	1900.2	1069.6	13277.5	3588.5
## 2	286.4	2.2	563.9	203.5
## 3	38.7	4.3	167.5	59.6
## 4	2.5	NA	0.2	0.2
## 5	94.8	7.4	349.7	215.8
## 6	864.9	22.7	1296.2	278.5
##	Quick.ratio..times.	Current.ratio..times.	Debt.to.equity.ratio..times.	
## 1	1.18	1.37		0.00
## 2	0.95	1.56		0.78
## 3	1.11	1.55		0.35
## 4	NA	NA		0.00
## 5	1.41	2.54		1.79
## 6	0.48	1.27		1.09
##	Cash.to.current.liabilities..times.			
## 1		0.43		
## 2		0.06		
## 3		0.21		
## 4		NA		
## 5		0.00		
## 6		0.11		
##	Cash.to.average.cost.of.sales.per.day	Creditors.turnover		
## 1		68.21	3.62	
## 2		5.96	9.80	
## 3		17.07	5.28	
## 4		NA	0.00	
## 5		0.00	13.00	
## 6		15.78	6.50	
##	Debtors.turnover	Finished.goods.turnover	WIP.turnover	
## 1	3.85	200.55	21.78	
## 2	5.70	14.21	7.49	
## 3	5.07	9.24	0.23	
## 4	0.00	NA	NA	
## 5	9.46	12.68	7.90	
## 6	21.13	10.14	8.38	
##	Raw.material.turnover	Shares.outstanding	Equity.face.value	EPS
## 1	7.71	42381675	10	35.52
## 2	11.46	11550000	10	9.97
## 3	NA	8149090	10	-0.50
## 4	0.00	52404	10	0.00
## 5	17.03	619635	10	7.91
## 6	4.74	1141718	10	30.57
##	Adjusted.EPS	Total.liabilities	PE.on.BSE	
## 1	7.10	17512.3	27.31	
## 2	9.97	941.0	8.17	
## 3	-0.50	232.8	-5.76	
## 4	0.00	2.7	NA	
## 5	7.91	478.5	NA	
## 6	15.28	2434.4	NA	

```
Default=ifelse(raw_data$Networth.Next.Year>0,0,1)
```

```
summary(as.factor(Default))
```

```
##      0      1  
## 3298  243
```

### This means that there are 3298 non defaulting Companies whereas 243 defaulting Companies in the given dataset.

```
val =read.csv('validation_data.csv')
```

```
str(raw_data)
```

```
## 'data.frame':    3541 obs. of  52 variables:  
## $ Num                               : int  1 2 3 4 5 6 7 8 9 10 ...  
## $ Networth.Next.Year                 : num  8890.6 394.3 92.2 2.7 109 ...  
## $ Total.assets                       : num  17512.3 941 232.8 2.7 478.5  
...  
## $ Net.worth                         : num  7093.2 351.5 100.6 2.7 107.6  
...  
## $ Total.income                      : num  24965 1527 477 NA 1580 ...  
## $ Change.in.stock                   : num  235.8 42.7 -5.2 NA -17 ...  
## $ Total.expenses                    : num  23658 1455 479 NA 1558 ...  
## $ Profit.after.tax                  : num  1543.2 115.2 -6.6 NA 5.5 ...  
## $ PBDITA                           : num  2860.2 283 5.8 NA 31 ...  
## $ PBT                              : num  2417.2 188.4 -6.6 NA 6.3 ...  
## $ Cash.profit                      : num  1872.8 158.6 0.3 NA 11.9 ...  
## $ PBDITA.as...of.total.income       : num  11.46 18.53 1.22 0 1.96 ...  
## $ PBT.as...of.total.income          : num  9.68 12.33 -1.38 0 0.4 ...  
## $ PAT.as...of.total.income          : num  6.18 7.54 -1.38 0 0.35 2.81 0  
0.72 8.29 -2.88 ...  
## $ Cash.profit.as...of.total.income  : num  7.5 10.38 0.06 0 0.75 ...  
## $ PAT.as...of.net.worth             : num  23.78 38.08 -6.35 0 5.25 ...  
## $ Sales                            : num  24458 1504 476 NA 1575 ...  
## $ Income.from.financial.services    : num  158 4 1.5 NA 3.9 6.4 NA NA  
7.3 NA ...  
## $ Other.income                     : num  297.2 15.9 0.2 NA 0.9 ...  
## $ Total.capital                    : num  423.8 115.5 81.4 0.5 6.2 ...  
## $ Reserves.and.funds               : num  6822.8 257.8 19.2 2.2 161.8  
...  
## $ Deposits..accepted.by.commercial.banks. : logi  NA NA NA NA NA NA ...  
## $ Borrowings                       : num  14.9 272.5 35.4 NA 193.1 ...  
## $ Current.liabilities...provisions    : num  9965.9 210 96.8 NA 112.8 ...  
## $ Deferred.tax.liability            : num  284.9 85.2 NA NA 4.6 ...  
## $ Shareholders.funds                : num  7093.2 351.5 100.6 2.7 107.6  
...  
## $ Cumulative.retained.profits        : num  6263.3 247.4 32.4 2.2 82.7  
...  
## $ Capital.employed                  : num  7108.1 624 136 2.7 300.7 ...  
## $ TOL.TNW                          : num  1.33 1.23 1.44 0 2.83 1.8  
0.03 5.17 1.05 3.25 ...  
## $ Total.term.liabilities...tangible.net.worth: num  0 0.34 0.29 0 1.59 0.37 0.03
```

```

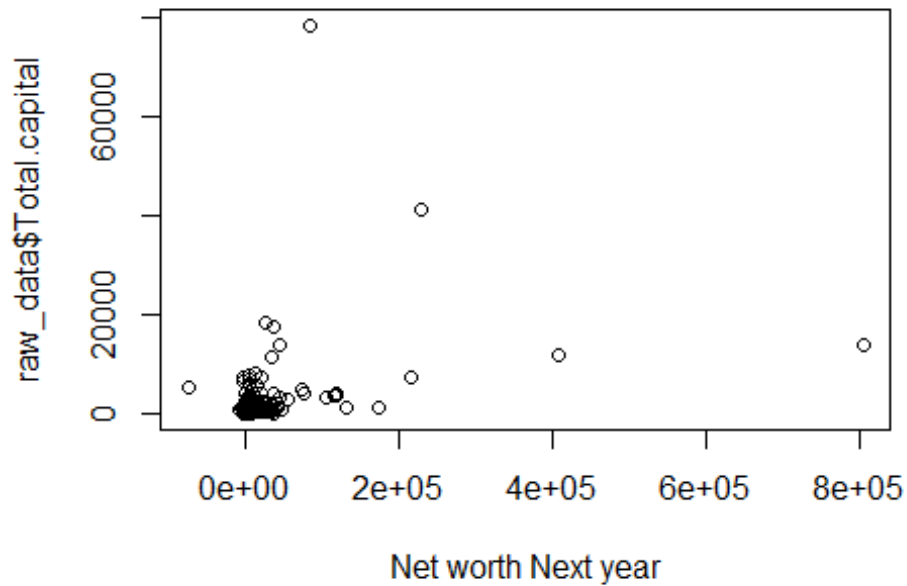
0.94 0.3 0.54 ...
## $ Contingent.liabilities...Net.worth.... : num 14.8 19.2 45.8 0 34.9 ...
## $ Contingent.liabilities : num 1049.7 67.6 46.1 NA 37.6 ...
## $ Net.fixed.assets : num 1900.2 286.4 38.7 2.5 94.8
...
## $ Investments : num 1069.6 2.2 4.3 NA 7.4 ...
## $ Current.assets : num 13277.5 563.9 167.5 0.2 349.7
...
## $ Net.working.capital : num 3588.5 203.5 59.6 0.2 215.8
...
## $ Quick.ratio..times. : num 1.18 0.95 1.11 NA 1.41 0.48
NA 0.54 0.59 0.39 ...
## $ Current.ratio..times. : num 1.37 1.56 1.55 NA 2.54 1.27
NA 1.15 1.58 0.5 ...
## $ Debt.to.equity.ratio..times. : num 0 0.78 0.35 0 1.79 1.09 0.32
2.31 0.94 3.13 ...
## $ Cash.to.current.liabilities..times. : num 0.43 0.06 0.21 NA 0 0.11 NA
0.04 0.19 0 ...
## $ Cash.to.average.cost.of.sales.per.day : num 68.21 5.96 17.07 NA 0 ...
## $ Creditors.turnover : num 3.62 9.8 5.28 0 13 ...
## $ Debtors.turnover : num 3.85 5.7 5.07 0 9.46 ...
## $ Finished.goods.turnover : num 200.55 14.21 9.24 NA 12.68
...
## $ WIP.turnover : num 21.78 7.49 0.23 NA 7.9 ...
## $ Raw.material.turnover : num 7.71 11.46 NA 0 17.03 ...
## $ Shares.outstanding : num 42381675 11550000 8149090
52404 619635 ...
## $ Equity.face.value : num 10 10 10 10 10 10 10 NA 10 10
...
## $ EPS : num 35.52 9.97 -0.5 0 7.91 ...
## $ Adjusted.EPS : num 7.1 9.97 -0.5 0 7.91 ...
## $ Total.liabilities : num 17512.3 941 232.8 2.7 478.5
...
## $ PE.on.BSE : num 27.31 8.17 -5.76 NA NA ...

## Univariate Analysis

plot(raw_data$Networth.Next.Year,raw_data$Total.capital, main="Histogram of Networth
Next Year",
xlab="Net worth Next year")

```

## Histogram of Network Next Year



```
raw1 =raw_data[,c(3:52)]
raw1 =cbind(Default,raw1)
raw1[,c(2:51)] =sapply(raw1[,c(2:51)], function(x)
  (x-mean(x[!is.na(x)]))/sd(x[!is.na(x)]))
colSums(is.na(raw1))

##              Default
##              0
##      Total.assets
##              0
##      Net.worth
##              0
##      Total.income
##              198
##      Change.in.stock
##              458
##      Total.expenses
##              139
##      Profit.after.tax
##              131
##      PBDITA
##              131
##      PBT
##              131
##      Cash.profit
```

##		131
##	PBDITA.as...of.total.income	
##		68
##	PBT.as...of.total.income	
##		68
##	PAT.as...of.total.income	
##		68
##	Cash.profit.as...of.total.income	
##		68
##	PAT.as...of.net.worth	
##		0
##	Sales	
##		259
##	Income.from.financial.services	
##		935
##	Other.income	
##		1295
##	Total.capital	
##		4
##	Reserves.and.funds	
##		85
##	Deposits..accepted.by.commercial.banks.	
##		3541
##	Borrowings	
##		366
##	Current.liabilities...provisions	
##		96
##	Deferred.tax.liability	
##		1140
##	Shareholders.funds	
##		0
##	Cumulative.retained.profits	
##		38
##	Capital.employed	
##		0
##	TOL.TNW	
##		0
##	Total.term.liabilities...tangible.net.worth	
##		0
##	Contingent.liabilities...Net.worth....	
##		0
##	Contingent.liabilities	
##		1188
##	Net.fixed.assets	
##		118
##	Investments	
##		1435
##	Current.assets	
##		66
##	Net.working.capital	
##		32
##	Quick.ratio..times.	
##		93
##	Current.ratio..times.	
##		93

```

##          Debt.to.equity.ratio..times.
##          0
##      Cash.to.current.liabilities..times.
##          93
##      Cash.to.average.cost.of.sales.per.day
##          85
##          Creditors.turnover
##          333
##          Debtors.turnover
##          328
##          Finished.goods.turnover
##          740
##          WIP.turnover
##          640
##          Raw.material.turnover
##          361
##          Shares.outstanding
##          692
##          Equity.face.value
##          692
##          EPS
##          0
##          Adjusted.EPS
##          0
##          Total.liabilities
##          0
##          PE.on.BSE
##          2194

raw_imp =mice(raw1,m=5,maxit=20,defaultMethod ="pmm",seed=500)

## Warning: Number of logged events: 3506

raw2 =mice::complete(raw_imp,1)

raw2 =raw2[ , -which(names(raw2) %in%c("Deposits..accepted.by.commercial.banks."))]

raw2 =raw2[complete.cases(raw2), ]

write.csv(raw2,"raw.csv", row.names =FALSE)

val[,c(3:52)] =sapply(val[,c(3:52)], function(x)

  (x-mean(x[!is.na(x)]))/sd(x[!is.na(x)]))

val1 =val[,c(2:52)]

val_imp <-mice(val1,m=5,maxit=20,meth='pmm',seed=500)

## Warning: Number of logged events: 3506

val2 =mice::complete(val_imp,1)

val2 =val2[ , -which(names(val2) %in%c("Deposits..accepted.by.commercial.banks."))]

```

```

val2 =val2[complete.cases(val2), ]

write.csv(val2,"val_final.csv", row.names =FALSE)

full_model =glm(Default~.,data = raw2, family='binomial')

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

step_model =stepAIC(full_model, direction ='both')

raw3 =raw2[ , which(names(raw2)
%in%c("Default", "Total.assets", "Net.worth", "Total.income", "PBDITA", "PBT", "PAT.as...of
.total.income", "PAT.as...of.net.worth", "Sales", "Income.from.financial.services", "Tota
l.capital", "Reserves.and.funds", "Borrowings", "Current.liabilities...provisions", "Defe
rred.tax.liability", "TOL.TNW", "Total.term.liabilities...tangible.net.worth", "Net.fixe
d.assets", "Investments", "Quick.ratio..times.", "Current.ratio..times.", "Finished.goods
.turnover", "WIP.turnover", "PE.on.BSE"))]

val3 =val2[ , which(names(val2)
%in%c("Default...1", "Total.assets", "Net.worth", "Total.income", "PBDITA", "PBT", "PAT.as.
..of.total.income", "PAT.as...of.net.worth", "Sales", "Income.from.financial.services", "
Total.capital", "Reserves.and.funds", "Borrowings", "Current.liabilities...provisions", "
Deferred.tax.liability", "TOL.TNW", "Total.term.liabilities...tangible.net.worth", "Net.
fixed.assets", "Investments", "Quick.ratio..times.", "Current.ratio..times.", "Finished.g
oods.turnover", "WIP.turnover", "PE.on.BSE"))]

fit_model =glm(Default ~Total.assets +Net.worth +Total.income +PBDITA+
PBT +PAT.as...of.total.income +PAT.as...of.net.worth +
Sales +Income.from.financial.services +Total.capital +
Reserves.and.funds +Borrowings +Current.liabilities...provisions +
Deferred.tax.liability +TOL.TNW +Total.term.liabilities...tangible.net.worth +
Net.fixed.assets +Investments +Quick.ratio..times. +Current.ratio..times. +
Finished.goods.turnover +WIP.turnover +PE.on.BSE,data = raw2,family ='binomial')

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

vif(fit_model)

##
## Total.assets
## 544.405919
## Net.worth
## 20.636014
## Total.income
## 923.586623
## PBDITA
## 29.146225
## PBT
## 9.720235
## PAT.as...of.total.income
## 1.354144
## PAT.as...of.net.worth
## 1.233509
## Sales
## 798.882455

```

```

##          Income.from.financial.services
##                      4.549312
##                      Total.capital
##                      23.361141
##                      Reserves.and.funds
##                      12.862346
##                      Borrowings
##                      235.267617
##          Current.liabilities...provisions
##                      38.111351
##                      Deferred.tax.liability
##                      2.081500
##                      TOL.TNW
##                      14.504278
## Total.term.liabilities...tangible.net.worth
##                      14.548891
##                      Net.fixed.assets
##                      8.767035
##                      Investments
##                      1.627623
##                      Quick.ratio..times.
##                      5.308946
##                      Current.ratio..times.
##                      5.645261
##                      Finished.goods.turnover
##                      1.093182
##                      WIP.turnover
##                      1.125075
##                      PE.on.BSE
##                      1.007328

fit_model2 =glm(Default ~Net.worth +
PAT.as...of.total.income +PAT.as...of.net.worth +
Sales +Income.from.financial.services +Total.capital +
Borrowings +Current.liabilities...provisions +
Deferred.tax.liability+TOL.TNW +
Investments +Quick.ratio..times. +Current.ratio..times. +
Finished.goods.turnover +WIP.turnover +PE.on.BSE,data = raw2,family = 'binomial')

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

vif(fit_model2)

##          Net.worth          PAT.as...of.total.income
##          32.445241          1.514689
##          PAT.as...of.net.worth          Sales
##          1.092941          5.726098
##          Income.from.financial.services          Total.capital
##          5.478106          3.753918
##          Borrowings          Current.liabilities...provisions
##          12.834215          6.746077
##          Deferred.tax.liability          TOL.TNW
##          6.370873          1.066156
##          Investments          Quick.ratio..times.
##          6.017383          4.646934

```



```
##          Current.ratio..times.          Finished.goods.turnover
##                               4.818059                1.082121
##                               WIP.turnover                PE.on.BSE
##                               1.133133                1.008097
```

```
raw4 =raw2[ , which(names(raw2)
%in%c("Default","Net.worth","PAT.as...of.total.income","PAT.as...of.net.worth","Sales",
"Income.from.financial.services","Total.capital","Borrowings","Current.liabilities.
..provisions","Deferred.tax.liability","TOL.TNW","Net.fixed.assets","Investments","Qu
ick.ratio..times.","Current.ratio..times.","Finished.goods.turnover","WIP.turnover","
PE.on.BSE"))]
```

```
val4 =val2[ , which(names(val2)
%in%c("Default...1","Net.worth","PAT.as...of.total.income","PAT.as...of.net.worth","S
ales","Income.from.financial.services","Total.capital","Borrowings","Current.liabilit
ies...provisions","Deferred.tax.liability","TOL.TNW","Net.fixed.assets","Investments"
,"Quick.ratio..times.","Current.ratio..times.","Finished.goods.turnover","WIP.turnove
r","PE.on.BSE"))]
```

```
cor(raw4[2:18])
```

```
##          Net.worth PAT.as...of.total.income
## Net.worth          1.000000000          0.0077083835
## PAT.as...of.total.income          0.007708384          1.0000000000
## PAT.as...of.net.worth          0.017823649          0.0902314134
## Sales          0.796423550          0.0056371094
## Income.from.financial.services          0.580129939          -0.0395902875
## Total.capital          0.392756109          0.0042830397
## Borrowings          0.887319420          0.0071949997
## Current.liabilities...provisions          0.725949978          0.0060020957
## Deferred.tax.liability          0.952065008          0.0035787896
## TOL.TNW          -0.014096311          -0.0195192726
## Net.fixed.assets          0.972868326          0.0060136011
## Investments          0.880145668          0.0005368384
## Quick.ratio..times.          -0.006168826          -0.0301049746
## Current.ratio..times.          -0.006336952          -0.0243525908
## Finished.goods.turnover          -0.003600006          0.0001702175
## WIP.turnover          -0.008724276          -0.0256738344
## PE.on.BSE          -0.003838827          0.0037315711
##          PAT.as...of.net.worth          Sales
## Net.worth          0.017823649          0.796423550
## PAT.as...of.total.income          0.090231413          0.005637109
## PAT.as...of.net.worth          1.000000000          0.013298358
## Sales          0.013298358          1.000000000
## Income.from.financial.services          0.008372735          0.832404354
## Total.capital          0.003689746          0.288987946
## Borrowings          0.007663734          0.855990787
## Current.liabilities...provisions          0.020881326          0.827225004
## Deferred.tax.liability          0.007870282          0.851555060
## TOL.TNW          -0.089490986          -0.005992928
## Net.fixed.assets          0.011193485          0.787835262
## Investments          0.010182964          0.895556065
## Quick.ratio..times.          -0.011278507          -0.008042847
## Current.ratio..times.          -0.011470702          -0.007360653
## Finished.goods.turnover          0.029188896          -0.007420215
```

## WIP.turnover	-0.012792247	-0.004760999
## PE.on.BSE	0.010113966	-0.002647748
##	Income.from.financial.services	
## Net.worth	0.580129939	
## PAT.as...of.total.income	-0.039590288	
## PAT.as...of.net.worth	0.008372735	
## Sales	0.832404354	
## Income.from.financial.services	1.000000000	
## Total.capital	0.294010868	
## Borrowings	0.676620040	
## Current.liabilities...provisions	0.797331890	
## Deferred.tax.liability	0.664778258	
## TOL.TNW	-0.002184978	
## Net.fixed.assets	0.535537871	
## Investments	0.737360000	
## Quick.ratio..times.	0.006736201	
## Current.ratio..times.	0.007461886	
## Finished.goods.turnover	0.011323710	
## WIP.turnover	-0.006199066	
## PE.on.BSE	-0.003539680	
##	Total.capital	Borrowings
## Net.worth	0.392756109	0.887319420
## PAT.as...of.total.income	0.004283040	0.007195000
## PAT.as...of.net.worth	0.003689746	0.007663734
## Sales	0.288987946	0.855990787
## Income.from.financial.services	0.294010868	0.676620040
## Total.capital	1.000000000	0.329943252
## Borrowings	0.329943252	1.000000000
## Current.liabilities...provisions	0.316460268	0.745752999
## Deferred.tax.liability	0.397729188	0.898760246
## TOL.TNW	0.004182240	-0.001921664
## Net.fixed.assets	0.328944369	0.891819623
## Investments	0.218453216	0.895828058
## Quick.ratio..times.	-0.002222723	-0.009447987
## Current.ratio..times.	-0.004004821	-0.009107308
## Finished.goods.turnover	-0.001614468	0.014195664
## WIP.turnover	0.011535313	-0.003308503
## PE.on.BSE	-0.003915815	-0.004560432
##	Current.liabilities...provisions	
## Net.worth	0.7259499781	
## PAT.as...of.total.income	0.0060020957	
## PAT.as...of.net.worth	0.0208813265	
## Sales	0.8272250041	
## Income.from.financial.services	0.7973318899	
## Total.capital	0.3164602683	
## Borrowings	0.7457529990	
## Current.liabilities...provisions	1.0000000000	
## Deferred.tax.liability	0.8118119215	
## TOL.TNW	0.0027693617	
## Net.fixed.assets	0.7106331258	
## Investments	0.7501019888	
## Quick.ratio..times.	-0.0092861883	
## Current.ratio..times.	-0.0095206647	
## Finished.goods.turnover	-0.0073527972	
## WIP.turnover	-0.0002437027	

## PE.on.BSE	-0.0051645747	
##	Deferred.tax.liability	TOL.TNW
## Net.worth	0.952065008	-0.014096311
## PAT.as...of.total.income	0.003578790	-0.019519273
## PAT.as...of.net.worth	0.007870282	-0.089490986
## Sales	0.851555060	-0.005992928
## Income.from.financial.services	0.664778258	-0.002184978
## Total.capital	0.397729188	0.004182240
## Borrowings	0.898760246	-0.001921664
## Current.liabilities...provisions	0.811811922	0.002769362
## Deferred.tax.liability	1.000000000	0.014303166
## TOL.TNW	0.014303166	1.000000000
## Net.fixed.assets	0.955913828	-0.008161322
## Investments	0.859730788	-0.008446857
## Quick.ratio..times.	-0.008650166	-0.019583038
## Current.ratio..times.	-0.008510131	-0.017973348
## Finished.goods.turnover	-0.004767182	0.001889868
## WIP.turnover	-0.001296868	0.014430287
## PE.on.BSE	-0.004141897	-0.003438302
##	Net.fixed.assets	Investments
## Net.worth	0.972868326	0.8801456683
## PAT.as...of.total.income	0.006013601	0.0005368384
## PAT.as...of.net.worth	0.011193485	0.0101829636
## Sales	0.787835262	0.8955560647
## Income.from.financial.services	0.535537871	0.7373599997
## Total.capital	0.328944369	0.2184532155
## Borrowings	0.891819623	0.8958280577
## Current.liabilities...provisions	0.710633126	0.7501019888
## Deferred.tax.liability	0.955913828	0.8597307878
## TOL.TNW	-0.008161322	-0.0084468567
## Net.fixed.assets	1.000000000	0.8364733253
## Investments	0.836473325	1.0000000000
## Quick.ratio..times.	-0.008469338	-0.0066170654
## Current.ratio..times.	-0.008083859	-0.0065688691
## Finished.goods.turnover	-0.008227970	0.0085984485
## WIP.turnover	-0.004435649	0.0058152286
## PE.on.BSE	-0.003445401	-0.0037955947
##	Quick.ratio..times.	Current.ratio..times.
## Net.worth	-0.006168826	-0.006336952
## PAT.as...of.total.income	-0.030104975	-0.024352591
## PAT.as...of.net.worth	-0.011278507	-0.011470702
## Sales	-0.008042847	-0.007360653
## Income.from.financial.services	0.006736201	0.007461886
## Total.capital	-0.002222723	-0.004004821
## Borrowings	-0.009447987	-0.009107308
## Current.liabilities...provisions	-0.009286188	-0.009520665
## Deferred.tax.liability	-0.008650166	-0.008510131
## TOL.TNW	-0.019583038	-0.017973348
## Net.fixed.assets	-0.008469338	-0.008083859
## Investments	-0.006617065	-0.006568869
## Quick.ratio..times.	1.000000000	0.979788668
## Current.ratio..times.	0.979788668	1.000000000
## Finished.goods.turnover	0.047735906	0.048591950
## WIP.turnover	0.003207906	-0.003101069
## PE.on.BSE	0.003015215	0.008062199

```
## Finished.goods.turnover WIP.turnover
## Net.worth -0.0036000058 -0.0087242758
## PAT.as...of.total.income 0.0001702175 -0.0256738344
## PAT.as...of.net.worth 0.0291888962 -0.0127922474
## Sales -0.0074202145 -0.0047609992
## Income.from.financial.services 0.0113237096 -0.0061990662
## Total.capital -0.0016144676 0.0115353126
## Borrowings 0.0141956637 -0.0033085033
## Current.liabilities...provisions -0.0073527972 -0.0002437027
## Deferred.tax.liability -0.0047671821 -0.0012968682
## TOL.TNW 0.0018898680 0.0144302867
## Net.fixed.assets -0.0082279698 -0.0044356492
## Investments 0.0085984485 0.0058152286
## Quick.ratio..times. 0.0477359062 0.0032079062
## Current.ratio..times. 0.0485919501 -0.0031010685
## Finished.goods.turnover 1.0000000000 0.2406780827
## WIP.turnover 0.2406780827 1.0000000000
## PE.on.BSE -0.0059511197 -0.0070653213
## PE.on.BSE
## Net.worth -0.003838827
## PAT.as...of.total.income 0.003731571
## PAT.as...of.net.worth 0.010113966
## Sales -0.002647748
## Income.from.financial.services -0.003539680
## Total.capital -0.003915815
## Borrowings -0.004560432
## Current.liabilities...provisions -0.005164575
## Deferred.tax.liability -0.004141897
## TOL.TNW -0.003438302
## Net.fixed.assets -0.003445401
## Investments -0.003795595
## Quick.ratio..times. 0.003015215
## Current.ratio..times. 0.008062199
## Finished.goods.turnover -0.005951120
## WIP.turnover -0.007065321
## PE.on.BSE 1.000000000

fac_raw4 <-factanal(raw4[2:18],
factors =4, # number of factors to extract
rotation="promax",lower =0.1) # 'varimax' is an ortho rotation

fac_raw4
## Call:
## factanal(x = raw4[2:18], factors = 4, rotation = "promax", lower = 0.1)
##
## Uniquenesses:
## Net.worth PAT.as...of.total.income
## 0.100 0.997
## PAT.as...of.net.worth Sales
## 1.000 0.100
## Income.from.financial.services Total.capital
## 0.100 0.719
## Borrowings Current.liabilities...provisions
## 0.107 0.180
## Deferred.tax.liability TOL.TNW
```

```

##              0.100              0.999
##              Net.fixed.assets      Investments
##              0.100              0.100
##              Quick.ratio..times.    Current.ratio..times.
##              0.100              0.100
##              Finished.goods.turnover WIP.turnover
##              0.996              1.000
##              PE.on.BSE
##              1.000
##
## Loadings:
##              Factor1 Factor2 Factor3 Factor4
## Net.worth      0.988      -0.164
## PAT.as...of.total.income
## PAT.as...of.net.worth
## Sales          0.774      0.354
## Income.from.financial.services 0.352      0.719
## Total.capital                0.484
## Borrowings      0.974
## Current.liabilities...provisions 0.469      0.414      0.268
## Deferred.tax.liability 0.879      0.168
## TOL.TNW
## Net.fixed.assets 0.994      -0.215
## Investments      0.990      0.137      -0.218
## Quick.ratio..times.                0.969
## Current.ratio..times.                0.969
## Finished.goods.turnover
## WIP.turnover
## PE.on.BSE
##
##              Factor1 Factor2 Factor3 Factor4
## SS loadings    5.614   1.884   0.911   0.413
## Proportion Var 0.330   0.111   0.054   0.024
## Cumulative Var 0.330   0.441   0.495   0.519
##
## Factor Correlations:
##              Factor1 Factor2 Factor3 Factor4
## Factor1    1.0000 -0.01071 -0.44920  0.5551
## Factor2    -0.0107  1.00000 -0.00325 -0.0162
## Factor3    -0.4492 -0.00325  1.00000 -0.1914
## Factor4     0.5551 -0.01618 -0.19136  1.0000
##
## Test of the hypothesis that 4 factors are sufficient.
## The chi square statistic is 9234.7 on 74 degrees of freedom.
## The p-value is 0

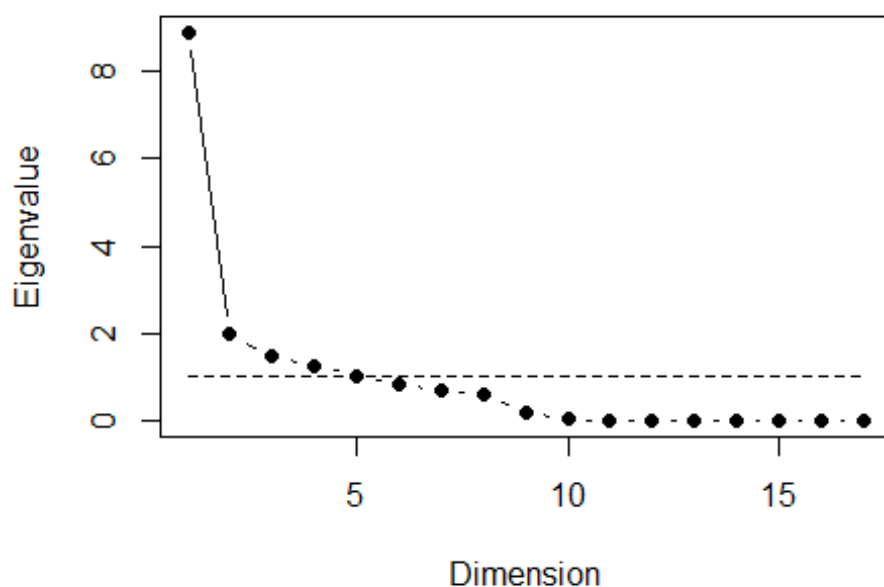
```

```

scree.plot(fac_raw4$correlation)

```

## Scree Plot



```
cor(raw4$Net.worth,raw4$Sales)

## [1] 0.7964235

f1_r
=raw4$Net.worth+raw4$Sales+raw4$Borrowings+raw4$Deferred.tax.liability+raw4$Net.fixed
.assets+raw4$Investments
f2_r =raw4$Quick.ratio..times.+raw4$Current.ratio..times.
f3_r =raw4$Income.from.financial.services
f4_r =raw4$Total.capital

raw_fac =data.frame("Default"=raw4$Default,"Company Size"
=f1_r,"Liquidity"=f4_r,"Profitability" =f3_r,"Leverage" =f2_r)

f1_v
=val4$Net.worth+val4$Sales+val4$Borrowings+val4$Deferred.tax.liability+val4$Net.fixed
.assets+val4$Investments
f2_v =val4$Quick.ratio..times.+val4$Current.ratio..times.
f3_v =val4$Income.from.financial.services
f4_v =val4$Total.capital

val_fac =data.frame("Default"=val4$Default...1,"Company Size"
=f1_v,"Liquidity"=f4_v,"Profitability" =f3_v,"Leverage" =f2_v)

log_fit_raw_fac =glm (Default~.,data = raw_fac,family ='binomial')

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

summary(log_fit_raw_fac)
```

```
##
## Call:
## glm(formula = Default ~ ., family = "binomial", data = raw_fac)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.8711  -0.3641  -0.3469  -0.3147   4.4825
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -3.07696    0.12082  -25.468  <2e-16 ***
## Company.Size  -0.11045    0.05779   -1.911   0.0560 .
## Liquidity      0.04406    0.04754    0.927   0.3540
## Profitability  0.39527    0.18980    2.083   0.0373 *
## Leverage     -1.65478    0.53802   -3.076   0.0021 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 1423.4  on 3281  degrees of freedom
## Residual deviance: 1402.5  on 3277  degrees of freedom
## AIC: 1412.5
##
## Number of Fisher Scoring iterations: 8

soft_pred =predict(log_fit_raw_fac,newdata = val_fac[2:5],type ='response')
hard_pred =as.factor(ifelse(soft_pred>0.075,1,0)) # Tried different values of p* to
get the value that gives best accuracy.
confusionMatrix(hard_pred,as.factor(val_fac$Default))

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    0    1
##              0 627  39
##              1   3   0
##
##              Accuracy : 0.9372
##              95% CI : (0.9161, 0.9544)
##      No Information Rate : 0.9417
##      P-Value [Acc > NIR] : 0.7236
##
##              Kappa : -0.0084
##
##      McNemar's Test P-Value : 6.641e-08
##
##              Sensitivity : 0.9952
##              Specificity : 0.0000
##      Pos Pred Value : 0.9414
##      Neg Pred Value : 0.0000
##      Prevalence : 0.9417
##      Detection Rate : 0.9372
##      Detection Prevalence : 0.9955
##      Balanced Accuracy : 0.4976
```

```
##
##      'Positive' Class : 0
##

val_pred = cbind(soft_pred, val_fac)

answer <- function(test){
  test2 = test[order(-test$soft_pred),]
  val
=c(sum(test2[1:67, "Default"]), sum(test2[68:134, "Default"]), sum(test2[135:201, "Default
"]),
sum(test2[202:268, "Default"]), sum(test2[269:335, "Default"]),
sum(test2[336:402, "Default"]), sum(test2[403:469, "Default"]),
sum(test2[470:536, "Default"]), sum(test2[537:603, "Default"]), sum(test2[604:669, "Defaul
t"])))
print(val)
}

answer(val_pred)

## [1] 13  8  3  4  3  3  2  0  3  0

### The accuracy derived is 93% due to class imbalance. The number above states that
13 out of the above class have been identified correctly.
```

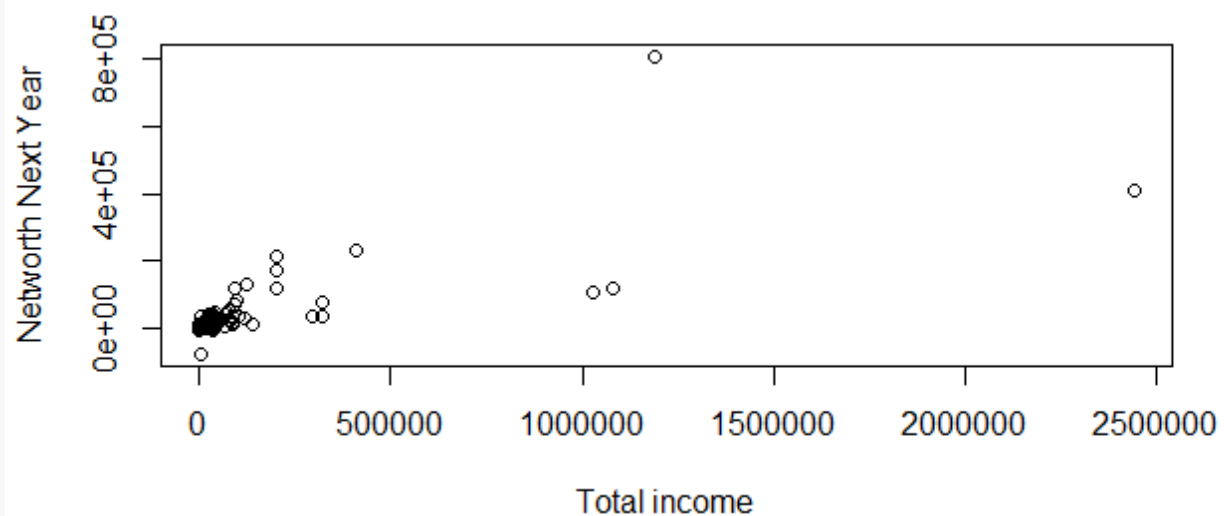


### OTHER OBSERVATIONS RELATED TO GIVEN DATASET

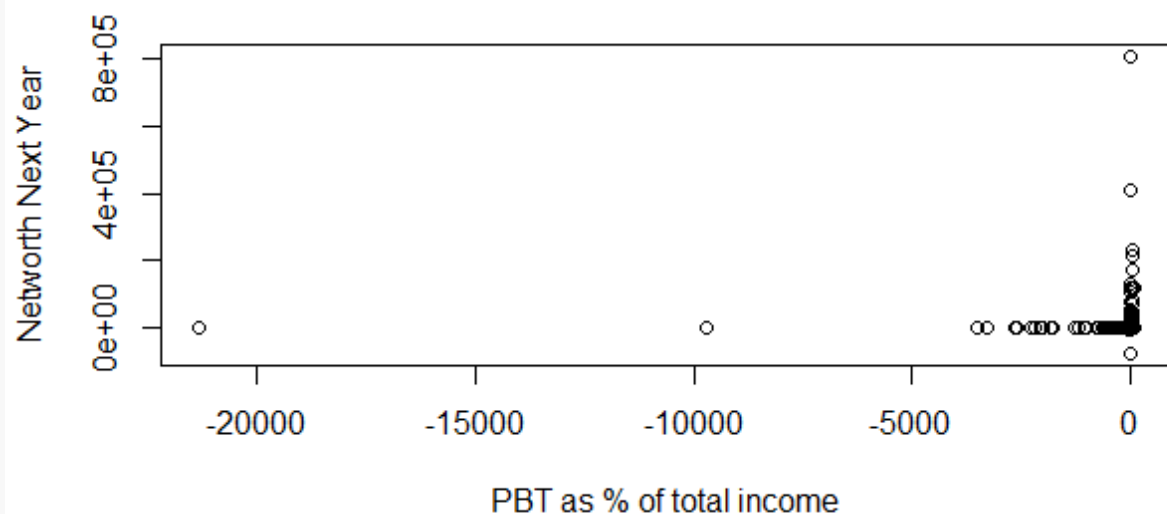
Based on Dr. Sarkar's Default risk estimation, following exercise was carried out for further understanding of dataset:

```
### Default rate
243/(3298+243)
### Default rate is 0.06862468
summary(Default)
summary(Default)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.00000 0.00000 0.00000 0.06862 0.00000 1.00000
```

```
plot(`Total income`, `Networth Next Year`)
```



```
plot(`PBT as % of total income`, `Networth Next Year`)
```



```
summary(`PBT as % of total income`)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
-21340.00	0.55	3.31	-17.28	8.80	100.00	68

### This shows that the min is -21340 which means that many companies in given dataset have a negative PBT over Total Income.

```
summary(`PBT as % of total income`[Default==0])
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
-9700.00	0.93	3.69	-3.97	9.29	100.00	52

### For companies whose default indicator is 0, i.e., companies who are non - Defaulters indicating a positive Net worth next year, median PBT as % Total Income Is 3.69. This speaks that a typical good company makes a profit of 3.69 units per 100 units of Income.

```
summary(`PBT as % of total income`[Default==1])
```

### However, For companies whose default indicator is 1, i.e., companies who are Defaulters indicating a negative Net worth next year, median PBT as % Total Income Is -5.93. It is evident that a bad company makes a loss of about 5.93 units per 100 units of Income.

### ### Logistic regression

#### ### Default Model 1 (PROFITABILITY)

```
Default.model1 = glm(as.factor(Default)~`PBT as % of total income`,  
                    family=binomial)
```

```
summary(Default.model1)
```

Call:

```
glm(formula = as.factor(Default) ~ `PBT as % of total income`,  
    family = binomial)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-4.1063	-0.3611	-0.3605	-0.3591	2.3921

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-2.696222	0.069883	-38.582	< 2e-16 ***
`PBT as % of total income`	-0.001147	0.000248	-4.626	3.72e-06 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1677.3 on 3472 degrees of freedom  
Residual deviance: 1647.0 on 3471 degrees of freedom  
(68 observations deleted due to missingness)  
AIC: 1651

Number of Fisher Scoring iterations: 5

### The analysis Of the summary clearly shows that the variable PBT as % Total Income is significant in impacting the dependant variable Net worth Next Year.

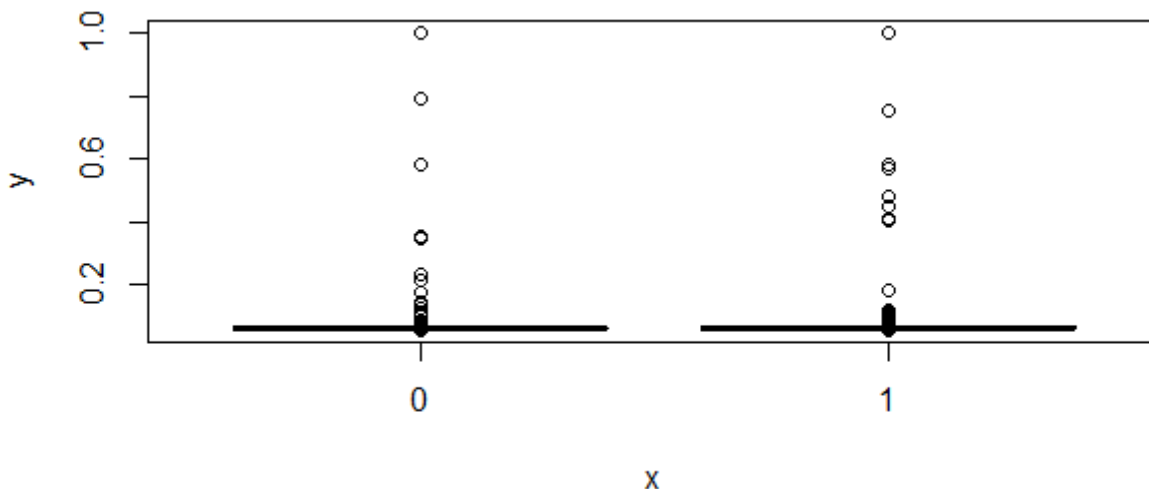
### Hence the equation derived through Logistic regression for the variable PBT as % of Total Income is significant:  
###  $-2.696222 + -0.001147(\text{PBT \% of Total Income})$

summary(Default.model1\$fitted.values)

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.05674	0.06260	0.06297	0.06536	0.06316	1.00000

### Fitted values refer to the probabilities of a company to do the default.  
### The summary says that the least risky company has a predicted default rate of 0.05674, a typical company has a median default rate of 0.06297, and  
### the worst of the companies show a very high predicted default rate of 1.

plot(as.factor(Default.model1\$y), Default.model1\$fitted.values)



### The plot reads that there is not a high discrimination between our actual probabilities and the fitted values. WWe need to build a richer model to understand the variables showing a reasonable discrimination between the actual probabilities and the fitted values being predicted probabilities.  
### Let us consider more variables and check the impact on the model.

### Default Model 2 (PROFITABILITY + LEVERAGE)

Default.model2=glm(as.factor(Default)~`PBT as % of total income`+  
`Debt to equity ratio (times)`, family=binomial)

summary(Default.model2)

Call:

glm(formula = as.factor(Default) ~ `PBT as % of total income` +  
`Debt to equity ratio (times)`, family = binomial)

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.1326	-0.3363	-0.3268	-0.3210	2.4836

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-2.9417330	0.0784617	-37.493	< 2e-16 ***
`PBT as % of total income`	-0.0010333	0.0002601	-3.973	7.10e-05 ***
`Debt to equity ratio (times)`	0.0637388	0.0083463	7.637	2.23e-14 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1677.3 on 3472 degrees of freedom  
Residual deviance: 1504.9 on 3470 degrees of freedom  
(68 observations deleted due to missingness)  
AIC: 1510.9

Number of Fisher Scoring iterations: 6

### Debt to equity Ratio as a variable is 7.637 standard deviations away from

### 0. Hence, this is even more significant than the PBT ratio as a % of Total

### Income.

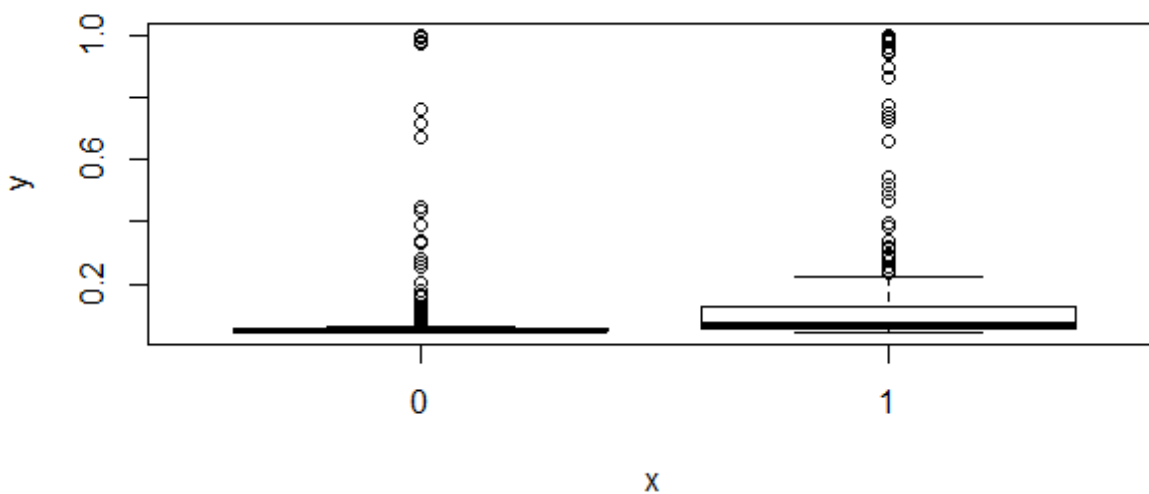
### Equation for this model is as follows:

###  $-2.9417330 - (0.0010333 * \text{PBT as \% of Total Income}) + (0.0637388 * \text{Debt to Equity Ratio})$

summary(Default.model2\$fitted.values)

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.04543	0.05061	0.05251	0.06536	0.05581	1.00000

plot(as.factor(Default.model2\$y), Default.model2\$fitted.values)



### It is seen that the 0s and 1s are separated more here. The upper Quartile

### of the predicted probabilities for the defaulting companies is more than

### 20% which is far higher than the range for the predicted probabilities for  
 ### non-defaulting companies. Hence, this model discriminates better between  
 ### good companies and not so good companies. Fitted default probabilities  
 ### is lower for good companies and higher for not so good companies.

### Let us convert this into an actual predictor of Default, i.e., when can we  
 ### say that a company has defaulted.  
 ### Based on our Default rate computation above  $(243/3298+243)$  - 6.86%, let us  
 ### analyse whether the knowledge of variables, profitability and leverage of  
 ### the company, whether it increases the rate of risk or decreases the rate  
 ### of risk.

```
Default.prediction2=ifelse(Default.model2$fitted.values>0.0686,1,0)
table(Default.model2$y, Default.prediction2)
Default.prediction2
      0      1
0 3113  133
1  107  120
```

### It says that we have been able to predict 3113 non-defaulters and 120  
 ### defaulters correctly, whereas, we have missed 107 defaulters to be  
 ### labelled correctly, and a 133 good companies that we have predicted as  
 ### default. The more riskier is that we have missed to be predict 107 bad  
 ### companies. Only 118 could be detected (half of the lot) and approx  
 ### other half -107 could not be detected.

### Let us loosen our ability to detect a risky company and play with the  
 ### default rate to reach an optimum find:

```
Default.prediction2=ifelse(Default.model2$fitted.values>0.06,1,0)
table(Default.model2$y, Default.prediction2)
Default.prediction2
      0      1
0 2929  317
1   81  146
```

```
146/(81+146)
[1] 0.6431718
```

### The sensitivity computed above shows that 64.31% of bad companies have  
 ### been identified.

### Let us see how good my prediction is for good companies:  
 $2929/(2929+317)$   
 0.9023413

### The specificity computed is 90.23% for good companies.  
 $(2929+146)/(2929+317+81+146)$   
 0.8854017

### The efficiency rate is 88.54%  
 $1-(2929+146)/(2929+317+81+146)$

0.1145983

### The rate of mis-classification is 11.45%. The mis-classification can be classified into following 2 categories:

- ### 1. Defaulters misclassified as Non-defaulters (Given loan might not come back)
- ### 2. Non-defaulters misclassified as Defaulters (Loan is not extended to such customers based on their mis-classified defaulting nature, hence, we lose out on prospective Income earning genres that were open to us - thereby placing hot cake like opportunities onto the lap of Competitors.)

### Choosing a good threshold depends on numerous factors other than the statistical knowhow and technique usage, such as, Business type, nature, statutory, legal, economical, political, social and geographical acumen. After taking into consideration all the pros and cons and regulatory compliances, then and only then, can a good threshold be defined and implemented through evaluations and elaboratively analysed and tested processes.

### For theoretical purposes, all things remaining constant, let us tighten the default rate as our base parameter:

```
Default.prediction2=ifelse(Default.model2$fitted.values>0.05,1,0)
```

```
table(Default.model2$y, Default.prediction2)
```

```
Default.prediction2
```

	0	1
0	436	2810
1	10	217

### 217 of companies have been correctly classified, however, essential numbers of good companies have been classified as defaulters. This ruins our Customer base.

```
Default.prediction2=ifelse(Default.model2$fitted.values>0.058,1,0)
```

```
table(Default.model2$y, Default.prediction2)
```

```
Default.prediction2
```

	0	1
0	2815	431
1	69	158

### Whether the table with default rate of 0.058 is better than the table for 0.06, is a business decision based on significant factors affecting such business. However, the universal principle of Investments states clearly that Reducing risk is directly related to reducing returns or opportunities.

```
rpart(as.factor(Default)~`PBT as % of total income`+
      `Debt to equity ratio (times)`)
```

n= 3541

```
node), split, n, loss, yval, (yprob)
```

\* denotes terminal node

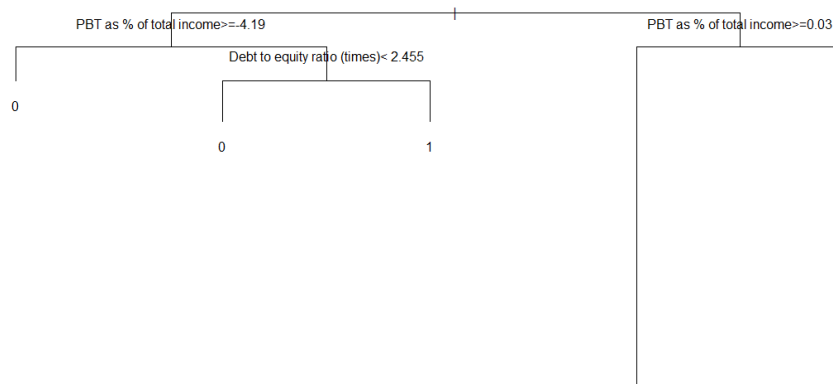
```

1) root 3541 243 0 (0.93137532 0.06862468)
  2) Debt to equity ratio (times)< 5.17 3307 127 0 (0.96159661 0.03840339)
    4) PBT as % of total income>=-4.19 3067 69 0 (0.97750245 0.02249755) *
    5) PBT as % of total income< -4.19 240 58 0 (0.75833333 0.24166667)
      10) Debt to equity ratio (times)< 2.455 198 33 0 (0.83333333
0.16666667) *
      11) Debt to equity ratio (times)>=2.455 42 17 1 (0.40476190
0.59523810) *
  3) Debt to equity ratio (times)>=5.17 234 116 0 (0.50427350 0.49572650)
    6) PBT as % of total income>=0.03 87 9 0 (0.89655172 0.10344828) *
    7) PBT as % of total income< 0.03 147 40 1 (0.27210884 0.72789116) *

```

```
plot(rpart(as.factor(Default)~`PBT as % of total income`+
`Debt to equity ratio (times)`))
```

```
text(rpart(as.factor(Default)~`PBT as % of total income`+
`Debt to equity ratio (times)`))
```



### ### Let us understand the Decision Tree. Out of 3541 companies, 243 had default.

### The first criteria asked was whether Debt Equity Ratio is more or less than

### 5.17. The default rate considered is 6.86%. If Debt equity ratio is less

### than 5.17, default rate plummets down to 3.84%, whereas, If Debt equity ratio ### is more than 5.17, default rate hikes up to 49.57%.

### Secondly, if companies are classified based on Debt equity ratio, the next

### criterion is to question whether the profitability is low or high.

### ### Such rules can be used as an alternative to logistic regression and

### Discriminant analysis.