

TP 3

TP 3

```
df <- read.table(file="../data/INFARCTUS.txt", header=TRUE, sep=" ", stringsAsFactors=FALSE)
head(df)
```

```
##   Obs C  PRONO FRCAR INCAR INSYS PRDIA PAPUL PVENT REPUL
## 1   1 2 SURVIE   90  1.71  19.0   16  19.5  16.0   912
## 2   2 1  DECES   90  1.68  18.7   24  31.0  14.0  1476
## 3   3 1  DECES  120  1.40  11.7   23  29.0   8.0  1657
## 4   4 2 SURVIE   82  1.79  21.8   14  17.5  10.0   782
## 5   5 1  DECES   80  1.58  19.7   21  28.0  18.5  1418
## 6   6 1  DECES   80  1.13  14.1   18  23.5   9.0  1664
```

1. ACP (selectionner les variables)

(a) Selection des variables

```
Xquantit <- df[, sapply(df, is.numeric)]
Xquantit
```

```
##   Obs C FRCAR INCAR INSYS PRDIA PAPUL PVENT REPUL
## 1   1 2   90  1.71  19.0  16.0  19.5  16.0   912
## 2   2 1   90  1.68  18.7  24.0  31.0  14.0  1476
## 3   3 1  120  1.40  11.7  23.0  29.0   8.0  1657
## 4   4 2   82  1.79  21.8  14.0  17.5  10.0   782
## 5   5 1   80  1.58  19.7  21.0  28.0  18.5  1418
## 6   6 1   80  1.13  14.1  18.0  23.5   9.0  1664
## 7   7 2   94  2.04  21.7  23.0  27.0  10.0  1059
## 8   8 2   80  1.19  14.9  16.0  21.0  16.5  1412
## 9   9 2   78  2.16  27.7  15.0  20.5  11.5   759
## 10  10 2  100  2.28  22.8  16.0  23.0   4.0   807
## 11  11 2   90  2.79  31.0  16.0  25.0   8.0   717
## 12  12 2   86  2.70  31.4  15.0  23.0   9.5   681
## 13  13 2   80  2.61  32.6   8.0  15.0   1.0   460
## 14  14 2   61  2.84  47.3  11.0  17.0  12.0   479
## 15  15 2   99  3.12  31.8  15.0  20.0  11.0   513
## 16  16 2   92  2.47  26.8  12.0  19.0  11.0   615
## 17  17 2   96  1.88  19.6  12.0  19.0   3.0   809
## 18  18 2   86  1.70  19.8  10.0  14.0  10.5   659
## 19  19 2  125  3.37  26.9  18.0  28.0   6.0   665
## 20  20 2   80  2.01  25.0  15.0  20.0   6.0   796
## 21  21 2   82  3.15  38.4  13.0  20.0   6.0   508
```

## 22	22 1	110	1.66	15.1	23.0	31.0	6.5	1494
## 23	23 1	80	1.50	18.7	13.0	17.0	12.0	907
## 24	24 1	118	1.03	8.7	19.0	27.0	10.0	2097
## 25	25 1	95	1.89	19.9	25.0	27.0	20.0	1143
## 26	26 1	80	1.45	18.1	19.0	23.0	15.0	1269
## 27	27 1	85	1.30	15.1	13.0	18.0	10.0	1108
## 28	28 1	105	1.84	17.5	18.0	22.0	10.0	957
## 29	29 2	122	2.79	22.9	25.0	36.0	10.0	1032
## 30	30 2	81	1.77	21.9	18.0	27.0	11.0	1220
## 31	31 2	118	2.31	19.6	22.0	27.0	10.0	935
## 32	32 1	87	1.20	13.8	34.0	41.0	20.0	2733
## 33	33 1	65	1.19	18.3	15.0	18.0	13.0	1210
## 34	34 2	84	2.15	25.6	27.0	37.0	10.0	1377
## 35	35 1	103	0.91	8.8	30.0	33.5	10.0	2945
## 36	36 2	75	2.54	33.9	24.0	31.0	16.0	976
## 37	37 2	90	2.08	23.1	20.0	28.0	6.0	1077
## 38	38 2	90	1.93	21.4	11.0	18.0	10.0	746
## 39	39 1	90	0.95	10.6	20.0	24.0	6.0	2021
## 40	40 2	65	2.38	36.6	16.0	22.0	12.0	739
## 41	41 1	95	0.99	10.4	20.0	27.5	8.0	2222
## 42	42 1	95	0.85	8.9	19.0	22.0	15.5	2071
## 43	43 2	86	2.05	23.8	21.0	28.0	10.0	1093
## 44	44 2	82	2.02	24.6	16.0	22.0	14.0	871
## 45	45 1	70	1.44	20.6	19.0	26.5	11.0	1472
## 46	46 2	92	3.06	33.3	10.0	15.0	6.0	392
## 47	47 1	94	1.31	13.9	26.0	40.0	15.0	2443
## 48	48 1	79	1.29	16.3	24.0	31.0	10.0	1922
## 49	49 2	67	1.47	21.9	15.0	18.0	16.0	980
## 50	50 1	75	1.21	16.1	19.0	24.0	4.0	1587
## 51	51 2	80	2.41	30.9	19.0	24.0	7.0	797
## 52	52 2	61	3.28	54.0	12.0	16.0	7.0	390
## 53	53 1	110	1.24	11.3	22.0	27.5	11.0	1774
## 54	54 1	116	1.85	15.9	33.0	42.0	13.0	1816
## 55	55 2	75	2.00	26.7	16.0	22.0	5.0	880
## 56	56 1	92	1.97	21.4	18.0	27.0	3.0	1096
## 57	57 2	110	0.96	8.8	15.0	19.0	16.0	1583
## 58	58 2	95	2.56	26.9	8.0	13.0	3.0	406
## 59	59 2	75	2.32	30.9	8.0	10.0	6.0	345
## 60	60 2	80	2.65	33.1	13.0	19.0	9.0	574
## 61	61 1	102	1.60	15.7	24.0	31.0	16.0	1550
## 62	62 2	86	1.67	19.4	18.0	23.0	8.5	1102
## 63	63 1	60	0.82	13.7	22.0	32.0	13.0	3122
## 64	64 2	100	1.76	17.6	23.0	33.0	2.0	1500
## 65	65 2	80	3.28	41.0	12.0	17.0	2.0	415
## 66	66 2	108	2.96	27.4	24.0	35.0	6.5	946
## 67	67 1	92	1.37	14.8	25.0	46.0	11.0	2686
## 68	68 1	100	1.38	13.8	20.0	31.0	11.0	1797
## 69	69 2	80	2.85	35.6	25.0	32.0	7.0	898
## 70	70 1	87	2.51	28.8	16.0	24.0	20.0	765
## 71	71 2	100	2.31	23.1	8.0	12.0	1.0	416
## 72	72 1	120	1.18	9.9	25.0	36.0	8.0	2441
## 73	73 1	115	1.83	15.9	25.0	30.0	8.0	1311
## 74	74 2	101	2.55	25.2	23.2	30.5	9.0	957
## 75	75 2	92	2.17	23.5	19.0	24.0	3.0	885

```
## 76 76 1 87 1.42 16.1 20.0 26.0 10.0 1465
## 77 77 2 80 1.59 19.9 13.0 20.5 4.0 1031
## 78 78 1 88 1.47 16.7 23.0 32.5 10.0 1769
## 79 79 1 104 1.23 11.8 27.0 33.0 11.0 2146
## 80 80 2 90 1.45 16.1 17.0 24.0 8.5 1324
## 81 81 1 67 0.85 12.7 26.0 33.0 11.0 3106
## 82 82 2 87 2.37 27.2 15.0 22.0 10.0 743
## 83 83 2 108 2.40 22.2 26.0 31.0 4.0 1033
## 84 84 1 120 1.91 15.9 18.0 27.0 15.0 1131
## 85 85 1 108 1.50 13.9 28.0 43.0 16.0 1813
## 86 86 2 86 2.36 27.4 24.0 34.0 8.0 1153
## 87 87 1 112 1.56 13.9 24.0 29.0 4.0 1487
## 88 88 1 80 1.34 17.0 16.0 25.0 16.0 1493
## 89 89 1 95 1.65 17.4 20.0 33.0 7.0 1600
## 90 90 1 90 2.04 22.7 28.0 41.0 10.0 1608
## 91 91 2 90 3.03 33.6 17.0 23.5 7.0 620
## 92 92 1 94 1.21 12.9 17.0 22.0 3.0 1455
## 93 93 1 51 1.34 26.3 11.0 17.0 6.0 1015
## 94 94 1 110 1.17 10.6 29.0 35.0 10.5 2393
## 95 95 1 96 1.74 18.1 24.0 29.0 6.0 1333
## 96 96 1 132 1.31 9.9 23.0 28.0 12.0 1710
## 97 97 1 135 0.95 7.0 15.0 20.0 7.0 1684
## 98 98 1 105 1.92 18.3 18.0 24.0 3.0 1000
## 99 99 1 99 0.83 8.4 23.0 27.0 8.0 2602
## 100 100 1 116 0.60 5.2 33.0 38.0 10.0 5067
## 101 101 1 112 1.54 13.8 25.0 31.0 8.0 1610
```

```
Xquanti <- Xquanti[,-c(1,2)]
Xquanti
```

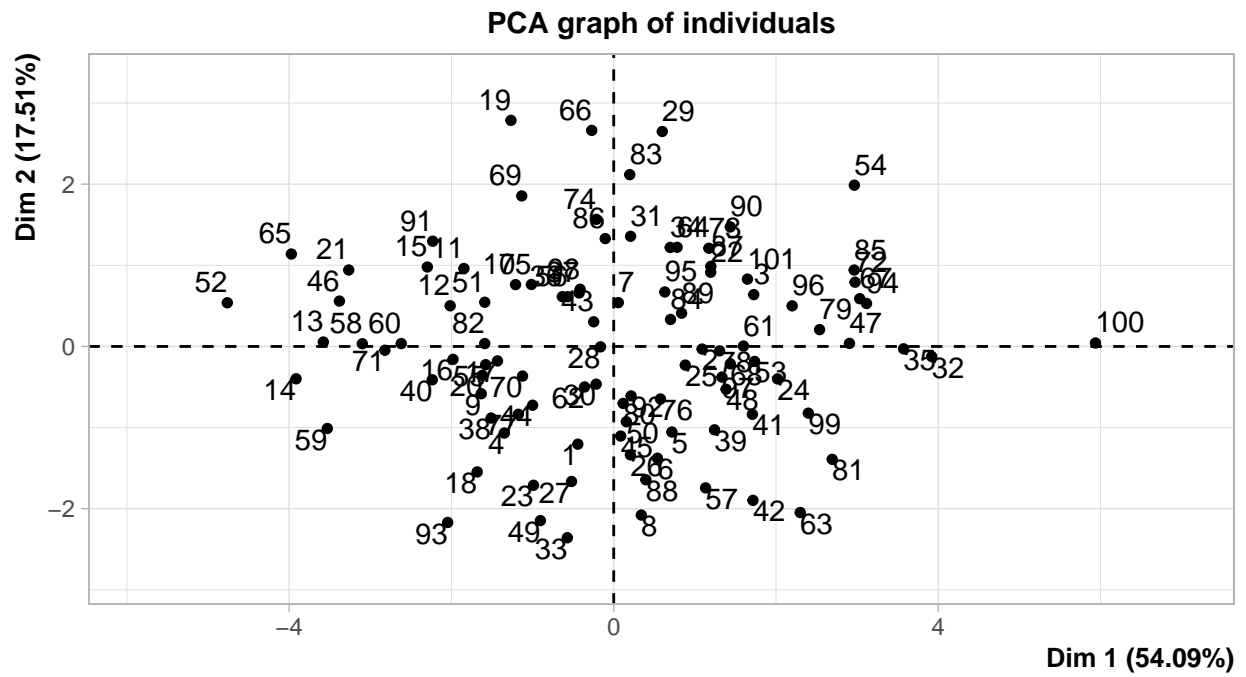
```
##      FRCAR INCAR INSYS PRDIA PAPUL PVENT REPUL
## 1      90  1.71  19.0  16.0  19.5  16.0  912
## 2      90  1.68  18.7  24.0  31.0  14.0  1476
## 3     120  1.40  11.7  23.0  29.0   8.0  1657
## 4      82  1.79  21.8  14.0  17.5  10.0   782
## 5      80  1.58  19.7  21.0  28.0  18.5  1418
## 6      80  1.13  14.1  18.0  23.5   9.0  1664
## 7      94  2.04  21.7  23.0  27.0  10.0  1059
## 8      80  1.19  14.9  16.0  21.0  16.5  1412
## 9      78  2.16  27.7  15.0  20.5  11.5   759
## 10     100  2.28  22.8  16.0  23.0   4.0   807
## 11      90  2.79  31.0  16.0  25.0   8.0   717
## 12      86  2.70  31.4  15.0  23.0   9.5   681
## 13      80  2.61  32.6   8.0  15.0   1.0   460
## 14      61  2.84  47.3  11.0  17.0  12.0   479
## 15      99  3.12  31.8  15.0  20.0  11.0   513
## 16      92  2.47  26.8  12.0  19.0  11.0   615
## 17      96  1.88  19.6  12.0  19.0   3.0   809
## 18      86  1.70  19.8  10.0  14.0  10.5   659
## 19     125  3.37  26.9  18.0  28.0   6.0   665
## 20      80  2.01  25.0  15.0  20.0   6.0   796
## 21      82  3.15  38.4  13.0  20.0   6.0   508
## 22     110  1.66  15.1  23.0  31.0   6.5  1494
## 23      80  1.50  18.7  13.0  17.0  12.0   907
```

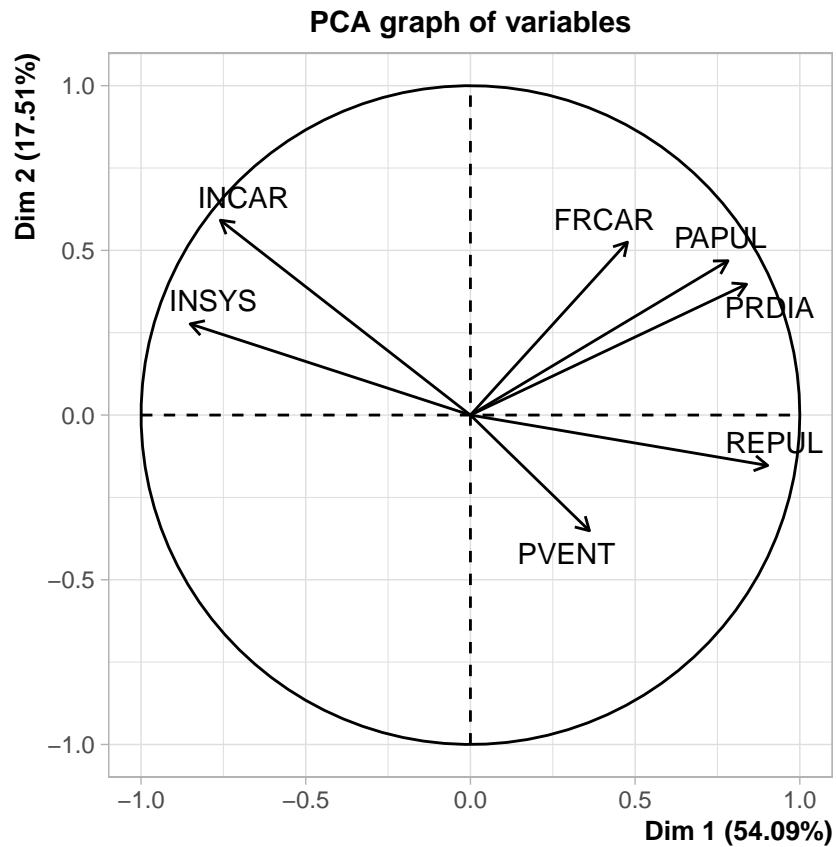
## 24	118	1.03	8.7	19.0	27.0	10.0	2097
## 25	95	1.89	19.9	25.0	27.0	20.0	1143
## 26	80	1.45	18.1	19.0	23.0	15.0	1269
## 27	85	1.30	15.1	13.0	18.0	10.0	1108
## 28	105	1.84	17.5	18.0	22.0	10.0	957
## 29	122	2.79	22.9	25.0	36.0	10.0	1032
## 30	81	1.77	21.9	18.0	27.0	11.0	1220
## 31	118	2.31	19.6	22.0	27.0	10.0	935
## 32	87	1.20	13.8	34.0	41.0	20.0	2733
## 33	65	1.19	18.3	15.0	18.0	13.0	1210
## 34	84	2.15	25.6	27.0	37.0	10.0	1377
## 35	103	0.91	8.8	30.0	33.5	10.0	2945
## 36	75	2.54	33.9	24.0	31.0	16.0	976
## 37	90	2.08	23.1	20.0	28.0	6.0	1077
## 38	90	1.93	21.4	11.0	18.0	10.0	746
## 39	90	0.95	10.6	20.0	24.0	6.0	2021
## 40	65	2.38	36.6	16.0	22.0	12.0	739
## 41	95	0.99	10.4	20.0	27.5	8.0	2222
## 42	95	0.85	8.9	19.0	22.0	15.5	2071
## 43	86	2.05	23.8	21.0	28.0	10.0	1093
## 44	82	2.02	24.6	16.0	22.0	14.0	871
## 45	70	1.44	20.6	19.0	26.5	11.0	1472
## 46	92	3.06	33.3	10.0	15.0	6.0	392
## 47	94	1.31	13.9	26.0	40.0	15.0	2443
## 48	79	1.29	16.3	24.0	31.0	10.0	1922
## 49	67	1.47	21.9	15.0	18.0	16.0	980
## 50	75	1.21	16.1	19.0	24.0	4.0	1587
## 51	80	2.41	30.9	19.0	24.0	7.0	797
## 52	61	3.28	54.0	12.0	16.0	7.0	390
## 53	110	1.24	11.3	22.0	27.5	11.0	1774
## 54	116	1.85	15.9	33.0	42.0	13.0	1816
## 55	75	2.00	26.7	16.0	22.0	5.0	880
## 56	92	1.97	21.4	18.0	27.0	3.0	1096
## 57	110	0.96	8.8	15.0	19.0	16.0	1583
## 58	95	2.56	26.9	8.0	13.0	3.0	406
## 59	75	2.32	30.9	8.0	10.0	6.0	345
## 60	80	2.65	33.1	13.0	19.0	9.0	574
## 61	102	1.60	15.7	24.0	31.0	16.0	1550
## 62	86	1.67	19.4	18.0	23.0	8.5	1102
## 63	60	0.82	13.7	22.0	32.0	13.0	3122
## 64	100	1.76	17.6	23.0	33.0	2.0	1500
## 65	80	3.28	41.0	12.0	17.0	2.0	415
## 66	108	2.96	27.4	24.0	35.0	6.5	946
## 67	92	1.37	14.8	25.0	46.0	11.0	2686
## 68	100	1.38	13.8	20.0	31.0	11.0	1797
## 69	80	2.85	35.6	25.0	32.0	7.0	898
## 70	87	2.51	28.8	16.0	24.0	20.0	765
## 71	100	2.31	23.1	8.0	12.0	1.0	416
## 72	120	1.18	9.9	25.0	36.0	8.0	2441
## 73	115	1.83	15.9	25.0	30.0	8.0	1311
## 74	101	2.55	25.2	23.2	30.5	9.0	957
## 75	92	2.17	23.5	19.0	24.0	3.0	885
## 76	87	1.42	16.1	20.0	26.0	10.0	1465
## 77	80	1.59	19.9	13.0	20.5	4.0	1031

```
## 78      88  1.47  16.7  23.0  32.5  10.0  1769
## 79     104  1.23  11.8  27.0  33.0  11.0  2146
## 80      90  1.45  16.1  17.0  24.0   8.5  1324
## 81      67  0.85  12.7  26.0  33.0  11.0  3106
## 82      87  2.37  27.2  15.0  22.0  10.0   743
## 83     108  2.40  22.2  26.0  31.0   4.0  1033
## 84     120  1.91  15.9  18.0  27.0  15.0  1131
## 85     108  1.50  13.9  28.0  43.0  16.0  1813
## 86      86  2.36  27.4  24.0  34.0   8.0  1153
## 87     112  1.56  13.9  24.0  29.0   4.0  1487
## 88      80  1.34  17.0  16.0  25.0  16.0  1493
## 89      95  1.65  17.4  20.0  33.0   7.0  1600
## 90      90  2.04  22.7  28.0  41.0  10.0  1608
## 91      90  3.03  33.6  17.0  23.5   7.0   620
## 92      94  1.21  12.9  17.0  22.0   3.0  1455
## 93      51  1.34  26.3  11.0  17.0   6.0  1015
## 94     110  1.17  10.6  29.0  35.0  10.5  2393
## 95      96  1.74  18.1  24.0  29.0   6.0  1333
## 96     132  1.31   9.9  23.0  28.0  12.0  1710
## 97     135  0.95   7.0  15.0  20.0   7.0  1684
## 98     105  1.92  18.3  18.0  24.0   3.0  1000
## 99      99  0.83   8.4  23.0  27.0   8.0  2602
## 100    116  0.60   5.2  33.0  38.0  10.0  5067
## 101    112  1.54  13.8  25.0  31.0   8.0  1610
```

```
library(FactoMineR)
```

```
res <- PCA(Xquanti)
```





2. Utiliser les qualitatives

```
X <- df[, -1]
X
```

##	C	PRONO	FRCAR	INCAR	INSYS	PRDIA	PAPUL	PVENT	REPUL
## 1	2	SURVIE	90	1.71	19.0	16.0	19.5	16.0	912
## 2	1	DECES	90	1.68	18.7	24.0	31.0	14.0	1476
## 3	1	DECES	120	1.40	11.7	23.0	29.0	8.0	1657
## 4	2	SURVIE	82	1.79	21.8	14.0	17.5	10.0	782
## 5	1	DECES	80	1.58	19.7	21.0	28.0	18.5	1418
## 6	1	DECES	80	1.13	14.1	18.0	23.5	9.0	1664
## 7	2	SURVIE	94	2.04	21.7	23.0	27.0	10.0	1059
## 8	2	SURVIE	80	1.19	14.9	16.0	21.0	16.5	1412
## 9	2	SURVIE	78	2.16	27.7	15.0	20.5	11.5	759
## 10	2	SURVIE	100	2.28	22.8	16.0	23.0	4.0	807
## 11	2	SURVIE	90	2.79	31.0	16.0	25.0	8.0	717
## 12	2	SURVIE	86	2.70	31.4	15.0	23.0	9.5	681
## 13	2	SURVIE	80	2.61	32.6	8.0	15.0	1.0	460
## 14	2	SURVIE	61	2.84	47.3	11.0	17.0	12.0	479
## 15	2	SURVIE	99	3.12	31.8	15.0	20.0	11.0	513
## 16	2	SURVIE	92	2.47	26.8	12.0	19.0	11.0	615
## 17	2	SURVIE	96	1.88	19.6	12.0	19.0	3.0	809
## 18	2	SURVIE	86	1.70	19.8	10.0	14.0	10.5	659

## 19	2	SURVIE	125	3.37	26.9	18.0	28.0	6.0	665
## 20	2	SURVIE	80	2.01	25.0	15.0	20.0	6.0	796
## 21	2	SURVIE	82	3.15	38.4	13.0	20.0	6.0	508
## 22	1	DECES	110	1.66	15.1	23.0	31.0	6.5	1494
## 23	1	DECES	80	1.50	18.7	13.0	17.0	12.0	907
## 24	1	DECES	118	1.03	8.7	19.0	27.0	10.0	2097
## 25	1	DECES	95	1.89	19.9	25.0	27.0	20.0	1143
## 26	1	DECES	80	1.45	18.1	19.0	23.0	15.0	1269
## 27	1	DECES	85	1.30	15.1	13.0	18.0	10.0	1108
## 28	1	DECES	105	1.84	17.5	18.0	22.0	10.0	957
## 29	2	SURVIE	122	2.79	22.9	25.0	36.0	10.0	1032
## 30	2	SURVIE	81	1.77	21.9	18.0	27.0	11.0	1220
## 31	2	SURVIE	118	2.31	19.6	22.0	27.0	10.0	935
## 32	1	DECES	87	1.20	13.8	34.0	41.0	20.0	2733
## 33	1	DECES	65	1.19	18.3	15.0	18.0	13.0	1210
## 34	2	SURVIE	84	2.15	25.6	27.0	37.0	10.0	1377
## 35	1	DECES	103	0.91	8.8	30.0	33.5	10.0	2945
## 36	2	SURVIE	75	2.54	33.9	24.0	31.0	16.0	976
## 37	2	SURVIE	90	2.08	23.1	20.0	28.0	6.0	1077
## 38	2	SURVIE	90	1.93	21.4	11.0	18.0	10.0	746
## 39	1	DECES	90	0.95	10.6	20.0	24.0	6.0	2021
## 40	2	SURVIE	65	2.38	36.6	16.0	22.0	12.0	739
## 41	1	DECES	95	0.99	10.4	20.0	27.5	8.0	2222
## 42	1	DECES	95	0.85	8.9	19.0	22.0	15.5	2071
## 43	2	SURVIE	86	2.05	23.8	21.0	28.0	10.0	1093
## 44	2	SURVIE	82	2.02	24.6	16.0	22.0	14.0	871
## 45	1	DECES	70	1.44	20.6	19.0	26.5	11.0	1472
## 46	2	SURVIE	92	3.06	33.3	10.0	15.0	6.0	392
## 47	1	DECES	94	1.31	13.9	26.0	40.0	15.0	2443
## 48	1	DECES	79	1.29	16.3	24.0	31.0	10.0	1922
## 49	2	SURVIE	67	1.47	21.9	15.0	18.0	16.0	980
## 50	1	DECES	75	1.21	16.1	19.0	24.0	4.0	1587
## 51	2	SURVIE	80	2.41	30.9	19.0	24.0	7.0	797
## 52	2	SURVIE	61	3.28	54.0	12.0	16.0	7.0	390
## 53	1	DECES	110	1.24	11.3	22.0	27.5	11.0	1774
## 54	1	DECES	116	1.85	15.9	33.0	42.0	13.0	1816
## 55	2	SURVIE	75	2.00	26.7	16.0	22.0	5.0	880
## 56	1	DECES	92	1.97	21.4	18.0	27.0	3.0	1096
## 57	2	SURVIE	110	0.96	8.8	15.0	19.0	16.0	1583
## 58	2	SURVIE	95	2.56	26.9	8.0	13.0	3.0	406
## 59	2	SURVIE	75	2.32	30.9	8.0	10.0	6.0	345
## 60	2	SURVIE	80	2.65	33.1	13.0	19.0	9.0	574
## 61	1	DECES	102	1.60	15.7	24.0	31.0	16.0	1550
## 62	2	SURVIE	86	1.67	19.4	18.0	23.0	8.5	1102
## 63	1	DECES	60	0.82	13.7	22.0	32.0	13.0	3122
## 64	2	SURVIE	100	1.76	17.6	23.0	33.0	2.0	1500
## 65	2	SURVIE	80	3.28	41.0	12.0	17.0	2.0	415
## 66	2	SURVIE	108	2.96	27.4	24.0	35.0	6.5	946
## 67	1	DECES	92	1.37	14.8	25.0	46.0	11.0	2686
## 68	1	DECES	100	1.38	13.8	20.0	31.0	11.0	1797
## 69	2	SURVIE	80	2.85	35.6	25.0	32.0	7.0	898
## 70	1	DECES	87	2.51	28.8	16.0	24.0	20.0	765
## 71	2	SURVIE	100	2.31	23.1	8.0	12.0	1.0	416
## 72	1	DECES	120	1.18	9.9	25.0	36.0	8.0	2441

##	73	1	DECES	115	1.83	15.9	25.0	30.0	8.0	1311
##	74	2	SURVIE	101	2.55	25.2	23.2	30.5	9.0	957
##	75	2	SURVIE	92	2.17	23.5	19.0	24.0	3.0	885
##	76	1	DECES	87	1.42	16.1	20.0	26.0	10.0	1465
##	77	2	SURVIE	80	1.59	19.9	13.0	20.5	4.0	1031
##	78	1	DECES	88	1.47	16.7	23.0	32.5	10.0	1769
##	79	1	DECES	104	1.23	11.8	27.0	33.0	11.0	2146
##	80	2	SURVIE	90	1.45	16.1	17.0	24.0	8.5	1324
##	81	1	DECES	67	0.85	12.7	26.0	33.0	11.0	3106
##	82	2	SURVIE	87	2.37	27.2	15.0	22.0	10.0	743
##	83	2	SURVIE	108	2.40	22.2	26.0	31.0	4.0	1033
##	84	1	DECES	120	1.91	15.9	18.0	27.0	15.0	1131
##	85	1	DECES	108	1.50	13.9	28.0	43.0	16.0	1813
##	86	2	SURVIE	86	2.36	27.4	24.0	34.0	8.0	1153
##	87	1	DECES	112	1.56	13.9	24.0	29.0	4.0	1487
##	88	1	DECES	80	1.34	17.0	16.0	25.0	16.0	1493
##	89	1	DECES	95	1.65	17.4	20.0	33.0	7.0	1600
##	90	1	DECES	90	2.04	22.7	28.0	41.0	10.0	1608
##	91	2	SURVIE	90	3.03	33.6	17.0	23.5	7.0	620
##	92	1	DECES	94	1.21	12.9	17.0	22.0	3.0	1455
##	93	1	DECES	51	1.34	26.3	11.0	17.0	6.0	1015
##	94	1	DECES	110	1.17	10.6	29.0	35.0	10.5	2393
##	95	1	DECES	96	1.74	18.1	24.0	29.0	6.0	1333
##	96	1	DECES	132	1.31	9.9	23.0	28.0	12.0	1710
##	97	1	DECES	135	0.95	7.0	15.0	20.0	7.0	1684
##	98	1	DECES	105	1.92	18.3	18.0	24.0	3.0	1000
##	99	1	DECES	99	0.83	8.4	23.0	27.0	8.0	2602
##	100	1	DECES	116	0.60	5.2	33.0	38.0	10.0	5067
##	101	1	DECES	112	1.54	13.8	25.0	31.0	8.0	1610

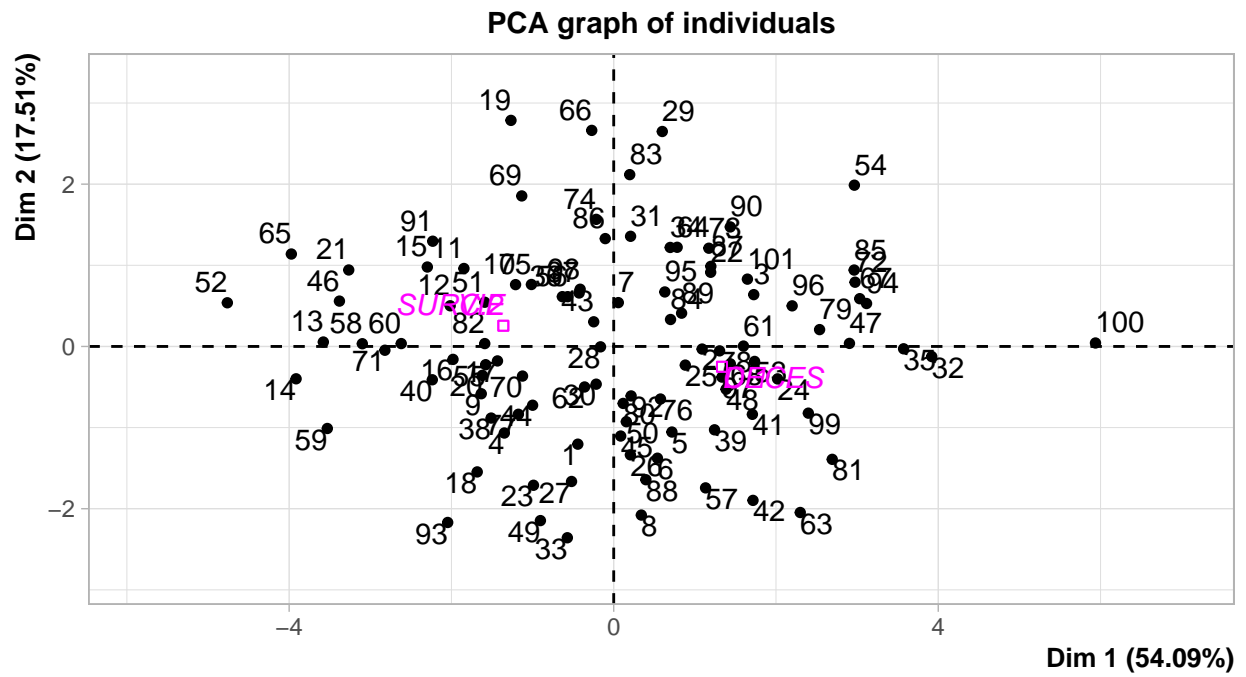
```
quali_indice <- c(1,2)
X[,quali_indice]
```

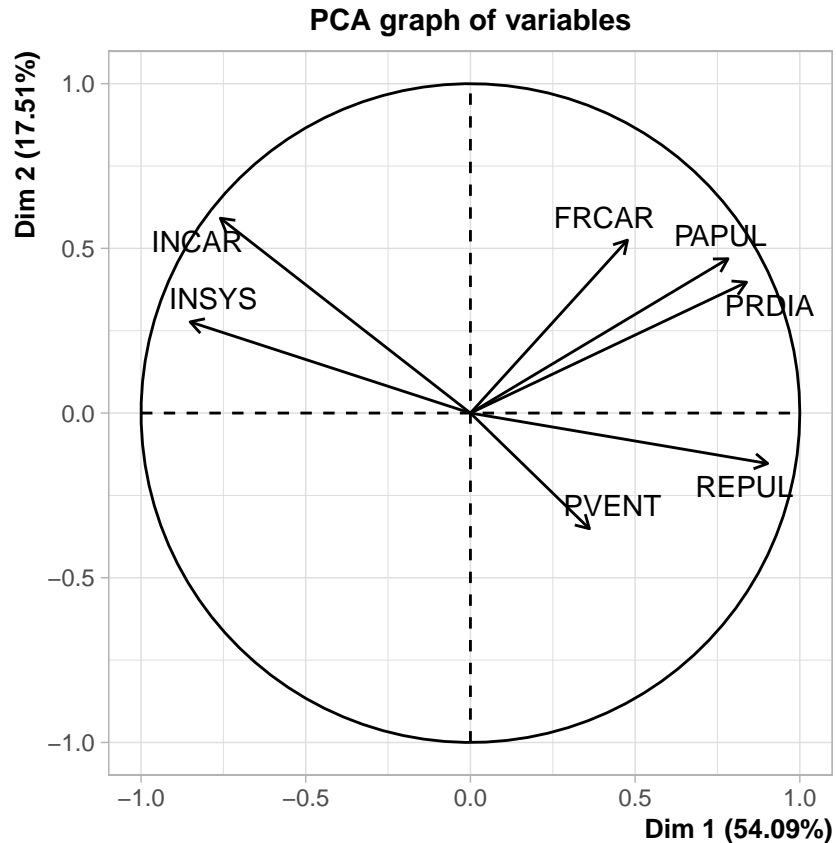
##		C	PRONO
##	1	2	SURVIE
##	2	1	DECES
##	3	1	DECES
##	4	2	SURVIE
##	5	1	DECES
##	6	1	DECES
##	7	2	SURVIE
##	8	2	SURVIE
##	9	2	SURVIE
##	10	2	SURVIE
##	11	2	SURVIE
##	12	2	SURVIE
##	13	2	SURVIE
##	14	2	SURVIE
##	15	2	SURVIE
##	16	2	SURVIE
##	17	2	SURVIE
##	18	2	SURVIE
##	19	2	SURVIE
##	20	2	SURVIE

21 2 SURVIE
22 1 DECES
23 1 DECES
24 1 DECES
25 1 DECES
26 1 DECES
27 1 DECES
28 1 DECES
29 2 SURVIE
30 2 SURVIE
31 2 SURVIE
32 1 DECES
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34 2 SURVIE
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36 2 SURVIE
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65 2 SURVIE
66 2 SURVIE
67 1 DECES
68 1 DECES
69 2 SURVIE
70 1 DECES
71 2 SURVIE
72 1 DECES
73 1 DECES
74 2 SURVIE

```
## 75 2 SURVIE
## 76 1 DECES
## 77 2 SURVIE
## 78 1 DECES
## 79 1 DECES
## 80 2 SURVIE
## 81 1 DECES
## 82 2 SURVIE
## 83 2 SURVIE
## 84 1 DECES
## 85 1 DECES
## 86 2 SURVIE
## 87 1 DECES
## 88 1 DECES
## 89 1 DECES
## 90 1 DECES
## 91 2 SURVIE
## 92 1 DECES
## 93 1 DECES
## 94 1 DECES
## 95 1 DECES
## 96 1 DECES
## 97 1 DECES
## 98 1 DECES
## 99 1 DECES
## 100 1 DECES
## 101 1 DECES
```

```
res_all <- PCA(
  X,
  quali.sup = quali_indice
)
```





3. Interprétation (10 lignes minimum). Quelles sont les variables liées au pronostic ?

Les variables qui jouent un rôle important sur le diagnostic sont les variables les mieux projetées sur le plan. Voici la liste des variables liées au pronostic :

- INCAR
- INSYS
- PAPUL
- PRDIA
- RPUL

Le nuage de points semble assez dispersé sur l'axe vertical. Et étalé sur l'axe horizontal avec une perte de densité sur les extrémités.

Les observations comprenant un décès ont tendance à être positives dans la première dimension tandis que celles qui comprennent une survie sont négatives sur la première dimension. La deuxième dimension paraît uniforme, donc sans conclusion.

En conclusion, les décès semblent être liés à un score positif sur les variables qui contribuent positivement à la première dimension (c'est-à-dire PAPUL, PRDIA, RPUL). Les variables contribuant le plus négativement à la première dimension sont plutôt liées à la survie (c'est-à-dire INCAR, INSYS);