Factor analysis of the athletics records data

Here we will fit one and two factor models to the athletics records data considered earlier; recall that these data consisted of national records (for men) in 8 running events from 100m to the marathon. A principal component analysis on the correlation matrix (that is, for standardized variables) revealed that the first two principal components had fairly simple interpretations.

The function factanal can be used to estimate the loadings and uniquenesses as well as giving an indication of the appropriateness of a given factor model. Maximum likelihood estimation is used and the variables are first standardized to have mean 0 and variance 1; this standardization is not necessary for factor analysis although it does ease interpretation of the loadings.

We will first try a one factor model:

```
> r0 <- factanal(~m100+m200+m400+m800+m1500+m5000+m10000+mara,factors=1)</pre>
> r0
Uniquenesses:
  m100
         m200
                m400
                        m800
                              m1500 m5000 m10000
                                                      mara
 0.447 0.389
                0.242
                       0.114 0.059
                                     0.104 0.369
                                                    0.205
Loadings:
       Factor1
m100
       0.743
m200
       0.782
m400
       0.871
m800
       0.941
m1500
      0.970
m5000
      0.946
m10000 0.794
mara
       0.892
                Factor1
                  6.070
SS loadings
                  0.759
Proportion Var
```

```
Test of the hypothesis that 1 factor is sufficient. The chi square statistic is 121.32 on 20 degrees of freedom. The p-value is 1.63e-16
```

The very small p-value strongly suggests that the one factor model is not adequate. Let's try a two factor model:

```
> r1 <- factanal(~m100+m200+m400+m800+m1500+m5000+m10000+mara,factors=2)</pre>
```

> r1

Uniquenesses:

```
m100 m200 m400 m800 m1500 m5000 m10000 mara 0.076 0.123 0.147 0.128 0.081 0.037 0.333 0.092
```

Loadings:

Factor1 Factor2

m100 0.295 0.915 m200 0.378 0.857

m400 0.552 0.740

m800 0.711 0.605

m1500 0.807 0.517

m5000 0.904 0.382

m10000 0.738 0.350

mara 0.913 0.271

Factor1 Factor2

SS loadings 3.887 3.095 Proportion Var 0.486 0.387

Cumulative Var 0.486 0.873

Test of the hypothesis that 2 factors are sufficient.

The chi square statistic is 15.03 on 13 degrees of freedom.

The p-value is 0.306

The p-value is much larger here than it is for the one factor model; this indicates that the two factor model fits the data quite well.

The function factanal uses the varimax rotation as its default. The loadings for the two factors here are essentially mirror images of each other, the first giving higher loadings to longer distances with the second giving higher loadings to shorter distances.

We will now try other rotations. First of all, we will set rotation="none", which applies no rotation.

```
> r2 < factanal(m100+m200+m400+m800+m1500+m5000+m10000+mara,factors=2,
```

+ rotation="none")

> r2

Uniquenesses:

m100 m200 m400 m800 m1500 m5000 m10000 mara 0.076 0.123 0.147 0.128 0.081 0.037 0.333 0.092

Loadings:

Factor1 Factor2

m100 0.787 0.552

m200 0.818 0.456

```
m400
        0.887
                0.258
m800
        0.932
m1500
        0.956
m5000
        0.952
              -0.240
m10000 0.800
              -0.165
        0.892
               -0.334
mara
               Factor1 Factor2
                 6.198
SS loadings
                          0.785
Proportion Var
                 0.775
                          0.098
Cumulative Var
                 0.775
                          0.873
```

Test of the hypothesis that 2 factors are sufficient.

The chi square statistic is 15.03 on 13 degrees of freedom.

The p-value is 0.306

Note that the loadings here are similar to the loadings for the first two principal components (although the signs and magnitudes are different.

Next we will apply the promax rotation (rotation="promax"). This is an oblique rotation (as opposed to an orthogonal rotation) and as a result, the two resulting factors are correlated.

```
> r3 <- factanal(~m100+m200+m400+m800+m1500+m5000+m10000+mara,factors=2,
+ rotation="promax")
> r3
```

Uniquenesses:

```
m100 m200 m400 m800 m1500 m5000 m10000 mara 0.076 0.123 0.147 0.128 0.081 0.037 0.333 0.092
```

Loadings:			
	Factor1	${\tt Factor2}$	
m100	-0.145	1.063	
m200		0.935	
m400	0.306	0.675	
m800	0.601	0.399	
m1500	0.783	0.223	
m5000	0.994		
m10000	0.788		
mara	1.072	-0.173	
		Factor1	Factor2
SS loadings		3.847	2.700
Proportion Var		0.481	0.337
Cumulative Var		0.481	0.818

Factor Correlations:

Factor1 Factor2

Factor1 1.000 0.735

Factor2 0.735 1.000

Test of the hypothesis that 2 factors are sufficient.

The chi square statistic is 15.03 on 13 degrees of freedom.

The p-value is 0.306

These loadings have a similar pattern to the loadings produced by the varimax rotation. Note that because the promax rotation is not an orthogonal rotation, the two factors are correlated.

Note that the overall fit of the two factor model does not depend on the rotation used; the idea behind rotating the loadings is to hopefully find factors (uncorrelated or not) whose loadings are easily interpretable.