

Same example as last time, but
start with R instead of S:

Stat 571
11-19-13
①

$$R = \begin{bmatrix} 1 & .4 & 0 \\ .4 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Note that R is
the same as the
covariance matrix
of the standardized
variables

$$\text{tr}(R) = 3$$

$$\begin{array}{ccc} \lambda_1 = 1.4 & \lambda_2 = 1 & \lambda_3 = .6 \\ \vec{e}_1 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix} & \vec{e}_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} & \vec{e}_3 = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \\ 0 \end{bmatrix} \end{array}$$

②

$$\begin{aligned} Y_1 &= .71 Z_1 + .71 Z_2 = .71 \left(\frac{X_1 - \mu_1}{16} \right) + .71 \left(\frac{X_2 - \mu_2}{10} \right) \\ &= .044(X_1 - \mu_1) + .071(X_2 - \mu_2) \end{aligned}$$

$$Y_2 = Z_3 = \frac{X_3 - \mu_3}{5}$$

etc.

	Z_1	Z_2	Z_3
Y_1	.84	.84	0
Y_2	0	0	1
Y_3	.55	-.55	0

Structure
matrix:

③ If we consider discarding some of the components, how many should go?

① Set a threshold for the cumulative percent variance.

Ex: Say we use 90% 279, 77, 25 out of 381

$$\text{Just } Y_1: \frac{279}{381} = .73$$

$$Y_1, Y_2: \frac{279+77}{381} = .93 \quad * \text{Stop}$$

Standardized: 1.4, 1, .6 out of 3

$$Y_1: 1.4/3 = .47$$

$$Y_1, Y_2, Y_3: 3/3 = 1 \quad *$$

$$Y_1, Y_2: 2.4/3 = .8$$

② Set a minimum eigenvalue
(only useful in the standardized version)

A commonly-used minimum is 1.

(called Kaiser's rule)

This discards any "below average" components.

In our example, Y_3 would be discarded.

③ If the original X_1, \dots, X_p are predictors in a multiple regression, use the p principal components as predictors instead, and keep the set of significant predictors.

Factor Analysis

③

Spearman & Pearson (1904)

		C	F	E	M	D	Mu
Classes	C	1	.83	.78	.70	.66	.63
French	F	.83	1	.67	.67	.65	.59
English	E	.78	.67	1	.64	.54	.51
Math	M	.70	.67	.64	1	.45	.51
Discern. Pitch	D	.66	.65	.54	.45	1	.40
Noise	Mu	.63	.59	.51	.51	.40	1

Look at the C & E columns +

⑥

notice: $\frac{.83}{.67} \approx \frac{.70}{.64} \approx \frac{.66}{.54} \approx \frac{.63}{.51}$

1.24 1.09 1.22 1.24

They searched for a model that would explain this behavior.

They proposed: $Z_i = G_i F + e_i$

\uparrow \uparrow \nwarrow
it *a single underlying*
standardized *factor, also standardized*
variable
0 expectation,
indep. of
F and
each
other

How does their model cause the constant ratio pattern that they observed?

(7)

$$E[z_i] = 0, \quad V[z_i] = 1$$

$$E[F] = 0, \quad E[F] = 1$$

$$V[z_i] = a_i^2 V[F] + V[e_i]$$

" "

1 1

$$1 - a_i^2 = V[e_i]$$

$$\text{Corr}[z_i, z_j] = \frac{\text{Cov}(z_i, z_j)}{\sqrt{V(z_i)V(z_j)}} = \text{Cov}(z_i, z_j)$$

$$= E[z_i z_j] - E[z_i]E[z_j]$$

$$= E[z_i z_j]$$

$$= E[(a_i F + e_i)(a_j F + e_j)]$$

$$= E[a_i a_j F^2 + a_i F e_j + a_j F e_i + e_i e_j]$$

$$= a_i a_j$$

$$\frac{\text{Cor}(z_i, z_j)}{\text{Cor}(z_i, z_k)} = \frac{a_i a_j}{a_i a_k} = \frac{a_j}{a_k}, \text{ which is the pattern they saw.}$$

(8)

⑨

$$z_i = a_i F + e_i$$

$$\frac{X_i - \mu_i}{\sigma_i} = a_i F + e_i$$

$$X_i = \mu_i + \sigma_i a_i F + \sigma_i e_i$$

$$X_i = \mu_i + \beta_i F + \varepsilon_i$$

Next: allow multiple factors

HW #6 due Nov 26 8.1, 8.2, 8.10 a,b ← do in a
file package