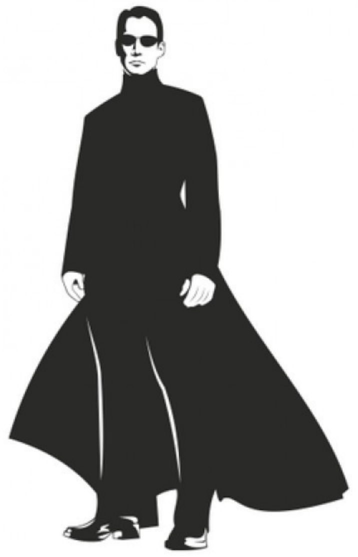
Matrix algebra

PSY544 – Introduction to Factor Analysis

Week 2



Prologue

- Matrix algebra is a framework for manipulating collections of numbers or algebraic symbols.

- Factor model is an algebraic system. If you understand the way it is communicated, you gain a better appreciation of what is going on.

- We have already seen the common factor model representing the structure of score x_{ij} – this model applies to every x_{ij} in the data matrix $\boldsymbol{X}$. Matrix algebra will allow us to express that.

Definitions

- **Scalar**: A single value, e.g., *k = 3, z = 0.7*
- **Matrix:** A rectangular table of *elements* (numbers, symbols…):

$$\boldsymbol{A} = \begin{bmatrix} & \end{bmatrix} \begin{bmatrix} & \end{bmatrix}$$

a_{ij} = element in row i and column j of matrix $\boldsymbol{A}$

$a_{11} = 1.0$, $a_{21} = 0.2$

An uppercase letter (like ***A***) names the matrix and stands for all the elements

Definitions

- **Data matrix:** Each element is a score for an individual on a variable

p = 3 manifest variables

$$\boldsymbol{X} = \begin{bmatrix} & & \\ & & \\ & & \\ & & \end{bmatrix}$$ N = 4 subjects

- x_{ij} = score for the *i*-th individual on the *j*-th variable

Definitions

- **Order**: The size of a matrix.
- A matrix with *N* rows and *p* columns is of order *N x p*

- **Square matrix**: A matrix with the same number of rows and columns
- **Vector**: A matrix with a single column (column vector) or a single row (row vector)

$$v = \begin{bmatrix} \\ \\ \\ \end{bmatrix}$$ m x 1 column vector

1 x m row vector

$$w = [\qquad\qquad\qquad]$$

Definitions

- **Transpose** of matrix ***A*** is a new matrix, ***A'***, formed by writing the rows of ***A*** as the columns of ***A'***:

If ***A =*** $\left[\qquad\qquad \right]$ then the transpose is ***A' =*** $\left[\qquad\qquad \right]$

- The transpose of a *P x M* matrix ***A*** is a *M x P* matrix ***A'*** (or $\boldsymbol{A}^{T}$)
- **Equality of matrices:** Two matrices are equal if and only if they are of the same order and all their corresponding elements are equal.

Arithmetic operations

- **Addition and subtraction:** Only possible if matrices are of the same order (*conformable* for addition)
- Corresponding elements of ***A*** and ***B*** are summed to form ***C***.
- ***C = A + B*** ; $c_{ij} = a_{ij} + b_{ij}$ for all *i* and *j*

$$\begin{bmatrix} \quad \end{bmatrix} \quad \begin{bmatrix} \quad \end{bmatrix} \quad \begin{bmatrix} \quad \end{bmatrix}$$

Commutative law: ***A + B = B + A***

Associative law: ***(A + B) + C = A + (B + C)***

The transpose of a sum is the sum of the transposes: ***(A + B)’ = A’ + B’***

Multiplication

- **Scalar multiplication**:

$$\boldsymbol{C} = \boldsymbol{B}a = a\boldsymbol{B}\ ;\ c_{ij} = a \cdot b_{ij}$$

$$2\left[\qquad\right] \quad \left[\qquad\right] \quad \left[\qquad\right]$$

Multiplication

- **Matrix multiplication** is **not** what you might think it is!
- **Matrix multiplication**: ***A*** and ***B*** have to be *conformable* for multiplication. Matrices are conformable if the number of columns in the first matrix equals the number of rows in the second.

$$\underset{n \times p}{\boldsymbol{A}} \quad \underset{p \times m}{\boldsymbol{B}} \quad = \boldsymbol{AB} = \quad \underset{n \times m}{\boldsymbol{C}}$$

- The product matrix has as many rows as the first matrix and as many columns as the second matrix.
- In this case, ***B*** was being *pre-multiplied* by ***A*** (and ***A*** was being *post-multiplied by B*)

Multiplication

- **Matrix multiplication**: c_{ij} equals the sum of products of elements in the *i*-th row of ***A*** and the *j*-th column of ***B***.

$$c_{ij} = a_{i1}\, b_{1j} + a_{i2}\, b_{2j} + \dots + a_{ip}\, b_{pj}$$

Multiplication

- For instance, if ***A*** is a 2x2 matrix and ***B*** is a 2x3 matrix, then ***AB = C*** is a 2x3 matrix:

$$\begin{array}{ccc} \boldsymbol{A} & \boldsymbol{B} & \boldsymbol{C} \\ \left[\quad\right] & \left[\qquad\right] & \left[\qquad\right] \\ \left[\quad\right] & \left[\qquad\right] & \left[\qquad\right] \end{array}$$

- Note again that order matters here! ***AB = C***, but ***BA*** is undefined.

Multiplication

- Matrix multiplication is associative: $\boldsymbol{A}(\boldsymbol{BC}) = \boldsymbol{AB}(\boldsymbol{C})$
- Matrix multiplication is not commutative: $\boldsymbol{AB} \neq \boldsymbol{BA}$
- Matrix multiplication is distributive: $\boldsymbol{A}(\boldsymbol{B}+\boldsymbol{C}) = \boldsymbol{AB} + \boldsymbol{AC}$; $(\boldsymbol{B}+\boldsymbol{C})\boldsymbol{A} = \boldsymbol{BA} + \boldsymbol{CA}$
- The transpose of a product of matrices equals the product of the transposes in reverse order: $(\boldsymbol{AB})' = \boldsymbol{B}' \boldsymbol{A}'$

Kinds of matrices

- **Symmetric matrix**: A square matrix ***S*** is symmetric if it's equal to its transpose

$$\boldsymbol{S} = \begin{bmatrix} & \\ & \end{bmatrix} = \boldsymbol{S}'$$

Kinds of matrices

- **Triangular matrix**: A square matrix ***S*** is lower triangular if all the elements above the diagonal are zero. A square matrix ***R*** is upper triangular if all the elements below the diagonal are equal to zero.

$$\boldsymbol{S} = \begin{bmatrix} \\ \\ \\ \end{bmatrix}$$

$$\boldsymbol{R} = \begin{bmatrix} \\ \\ \\ \end{bmatrix}$$

Kinds of matrices

- **Diagonal matrix**: A square matrix ***D*** is diagonal if all the off-diagonal elements are zero.

$$\boldsymbol{D} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

Kinds of matrices

- **Pre-multiplication** by a diagonal matrix scales rows:

$$\boldsymbol{DA} = \begin{bmatrix} \; \end{bmatrix}\begin{bmatrix} \; \end{bmatrix} \begin{bmatrix} \; \end{bmatrix}$$

- **Post-multiplication** by a diagonal matrix scales columns:

$$\boldsymbol{BD} = \begin{bmatrix} \; \end{bmatrix}\begin{bmatrix} \; \end{bmatrix} \begin{bmatrix} \; \end{bmatrix}$$

Kinds of matrices

- **Identity matrix** is a diagonal matrix with all diagonal elements = 1

$$\boldsymbol{I} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

- Multiplication of a matrix by an identity matrix does not change the matrix (it's like multiplying a scalar by 1)

Kinds of matrices

- **Orthogonal matrix**: A square matrix ***T*** is orthogonal if ***TT' = I*** or ***T'T = I***
- **Correlation matrix** is a square, symmetric matrix with unit diagonals and off-diagonal elements that satisfy $-1 \leq r_{ij} \leq 1$. Also, it has to be nonnegative definite (we will define that later)

Functions of matrices

- The **Determinant** of a square matrix $\boldsymbol{A}$ is a scalar function of the elements of $\boldsymbol{A}$. It is denoted as $|\boldsymbol{A}|$ or det($\boldsymbol{A}$) and is a single number (scalar).
- The determinant has many functions which we will not cover here (neither will we cover the definition or computation)
- If a matrix has determinant equal to zero, the matrix is called *singular*. This is an indication that there is redundancy among the rows / columns of the matrix – if the determinant is zero, some columns (or rows) of the matrix can be expressed as linear combinations of other columns (rows). In other words, the columns (rows) are linearly dependent.

Functions of matrices

- A singular matrix:

$$\boldsymbol{A} = \begin{bmatrix} \quad \end{bmatrix}$$

(The last column is the sum of the first two columns)

Functions of matrices

- **Trace:** The trace of a square matrix ***A***, tr(***A***), is the sum of its diagonal elements.

- **Rank:** The column rank of ***A*** is equal to the total number of linearly independent columns of ***A***. The row rank of ***A*** is equal to the total number of linearly independent rows of ***A***.

  The rank of an N x K matrix is at most the minimum of N or K, min(N,K)
- A matrix whose rank is equal to min(N,K) is *full rank*
- A matrix whose rank is less than min(N,K) is *rank deficient*

Functions of matrices

- **Inverse:** If $\boldsymbol{A}$ is a square matrix and is not singular (i.e., its determinant is non-zero), then it has a unique inverse $\boldsymbol{A^{-1}}$ such that:

$$\boldsymbol{AA^{-1} = A^{-1}A = I}$$

- The inverse of a matrix plays a role similar to that of a reciprocal in scalar algebra: $x * – = 1$
- Post-multiplying $\boldsymbol{A}$ by the inverse of $\boldsymbol{B}$ is analogous to „dividing“ $\boldsymbol{A}$ by $\boldsymbol{B}$ (assuming the matrices are conformable for multiplication)

Functions of matrices

- **Solving** equations:

Consider the equation $\boldsymbol{Ax = b}$, where $\boldsymbol{A}$ is a N x N non-singular matrix, $\boldsymbol{b}$ is a N x 1 vector and $\boldsymbol{x}$ is a N x 1 vector. We know the elements of $\boldsymbol{A}$ and $\boldsymbol{b}$ and wish to solve for $\boldsymbol{x}$:

$$\boldsymbol{Ax = b}$$

$$\boldsymbol{A^{-1}Ax = A^{-1}b}$$

$$\boldsymbol{Ix = A^{-1}b}$$

$$\boldsymbol{x = A^{-1}b}$$

Functions of matrices

- **Solving** equations:

$$4x + 5y = 4$$

$$3x + 1y = 3$$

- In matrix form:

[ ][ ] [ ]

solving:

[ ] [ ] [ ] [— — ; — —][ ] [ ]

Functions of matrices

- **Eigenvalues** and **Eigenvectors**
- Suppose that $\boldsymbol{S}$ is a square symmetric matrix of order p. If $\boldsymbol{u}$ is a column vector of order p and v is a scalar, such that:

$$\boldsymbol{Su} = v\boldsymbol{u}$$

...then v is said to be an *eigenvalue* (or characteristic root) of $\boldsymbol{S}$ and $\boldsymbol{u}$ is said to be an *eigenvector* (or characteristic vector) of $\boldsymbol{S}$.

- $\boldsymbol{S}$ will have p eigenvalues and p associated eigenvectors.

Functions of matrices

- **Eigenvalues** and **Eigenvectors**

- If all p eigenvalues are positive, the matrix is *positive definite*. If one or more eigenvalues are zero and the rest is positive, the matrix is *nonnegative definite*. If one or more eigenvalues are negative, the matrix is *negative definite*.

- The determinant of $\boldsymbol{S}$, det($\boldsymbol{S}$), equals the product of the eigenvalues of $\boldsymbol{S}$ Thus, if one or more eigenvalues are zero, the matrix is singular.

Functions of matrices

- **Eigenvalues** and **Eigenvectors**

- The eigenvalues can be arranged in descending order as the diagonal elements in a diagonal matrix ***D***, and the corresponding eigenvectors can be arranged as columns of matrix ***U***. Then:

  U is orthogonal, that is, ***U'U = I***
  The "eigenstructure" of ***S*** can be given in this form: ***SU = UD***
  It also holds that ***S = UDU'***

Linear combinations of random variables

- Matrix equations are handy for representing linear combinations of random variables
- Let $\boldsymbol{x}$ be a column vector of order p containing scores for a random individual on variables $x_1, x_2, \ldots, x_p$
- Let $\boldsymbol{z}$ be a column vector of order m containing scores for a random individual on variables $z_1, z_2, \ldots, z_m$
- We will represent the variables in $\boldsymbol{x}$ as linear functions of the variables in $\boldsymbol{z}$. Let $\boldsymbol{A}$ be a matrix of order p x m containing coefficients a_{jk} representing the linear effects of z_k on x_j
- Let $\boldsymbol{\mu}$ be a column vector of order p containing fixed constants $\mu_1, \mu_2, \ldots, \mu_p$

Linear combinations of random variables

- Then, we can represent the variables in ***x*** as linear functions of the variables in ***z*** and the constants in ***μ*** using the following matrix equation:

$$\boldsymbol{x} = \boldsymbol{\mu} + \boldsymbol{A}\boldsymbol{z}$$

…thus:

- The matrix equation actually contains the whole set of linear equations

An intermezzo – expected values

- Wiki: “The *expected value* of random variable is the long-run average value of repetitions of the experiment it represents”

...so, the *expected value* is the variable’s **mean**.

- $E[X] = \mu$

An intermezzo – expected values

- Now, consider the (scalar) formula for the **variance** of a random variable:

…which is the "mean squared deviation from the mean", right?

- As an expected value: $E[(X - \mu)^2]$

An intermezzo – expected values

- Now consider the (scalar) formula for **covariance**

…which is the "mean cross-product of deviations from the mean" (sorta)

- As an expected value: $E[(X - \mu_x)(Y - \mu_y)]$

Covariance matrix

- Now suppose that $\boldsymbol{x}$ is a vector of order p containing scores on p variables for a random individual selected from some population, and $\boldsymbol{\mu}$ is a vector of order p containing the population means of these p variables.
- Then, vector $(\boldsymbol{x} - \boldsymbol{\mu})$ stands for the vector $\boldsymbol{x}$ with the population means subtracted (it represents deviations from the mean)

- Let's multiply this vector by its transpose:

$$(\boldsymbol{x} - \boldsymbol{\mu})\,(\boldsymbol{x} - \boldsymbol{\mu})'$$

Covariance matrix

- Let's multiply this vector by its transpose:

$$(\boldsymbol{x} - \boldsymbol{\mu})\ (\boldsymbol{x} - \boldsymbol{\mu})'$$

- …and take the expectation:

$$\mathrm{E}[(\boldsymbol{x} - \boldsymbol{\mu})\ (\boldsymbol{x} - \boldsymbol{\mu})']$$

Covariance matrix

$$E[(\boldsymbol{x} - \boldsymbol{\mu})\,(\boldsymbol{x} - \boldsymbol{\mu})']$$

- Expanding, we get the expectation of:

$$\left[\quad\right]\left[\quad\right]$$

- …which gives us the variance/covariance matrix of the manifest variables

Covariance matrix

$$E[(\boldsymbol{x} - \boldsymbol{\mu})\,(\boldsymbol{x} - \boldsymbol{\mu})']$$

- The variance-covariance matrix is a *p* x *p* symmetric matrix with variances on the diagonal and covariances off the diagonal