



University of Ferrara

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# Matrix algebra for multivariate problems

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# Matrix algebra: linearly independent vectors

Vectors  $\mathbf{x}_1, \dots, \mathbf{x}_k$  are called **linearly dependent** if there exist numbers  $\lambda_1, \dots, \lambda_k$  not all zero such that

$$\lambda_1 \mathbf{x}_1 + \dots + \lambda_k \mathbf{x}_k = \mathbf{0}.$$

Otherwise the  $k$  vectors are linearly independent.

# Matrix algebra: linearly independent vectors

Let  $W$  be a subspace of  $\mathbf{R}^n$ .

Then a basis of  $W$  is a maximal linearly independent set of vectors.

Every basis of  $W$  contains the same (finite) number of elements.

This number is the dimension of  $W$ .

If  $\mathbf{x}_1, \dots, \mathbf{x}_k$  is a basis for  $W$  then every element  $\mathbf{x}$  in  $W$  can be expressed as a linear combination of  $\mathbf{x}_1, \dots, \mathbf{x}_k$ .

# Matrix algebra: linearly independent vectors

Example:

The dimension of  $\mathbf{W}=\mathbf{R}^3$  is 3.

A basis for  $\mathbf{R}^3$  is  $\mathbf{x}_1=(1,0,0)'$ ,  $\mathbf{x}_2=(0,1,0)'$  and  $\mathbf{x}_3=(0,0,1)'$ .

As a matter of fact  $\mathbf{x}_1$ ,  $\mathbf{x}_2$  and  $\mathbf{x}_3$  are linearly independent and every vector

$\mathbf{a}=(a_1, a_2, a_3)'$  can be expressed as linear combination of

$\mathbf{x}_1$ ,  $\mathbf{x}_2$  and  $\mathbf{x}_3$ :  $\mathbf{a} = a_1\mathbf{x}_1 + a_2\mathbf{x}_2 + a_3\mathbf{x}_3$

# Matrix algebra: the rank

The **rank** of a  $n \times k$  matrix  $\mathbf{A}$  is defined as the maximum number of linearly independent columns (rows) in  $\mathbf{A}$ .

The following properties hold for the rank of  $\mathbf{A}$ , denoted with  $r(\mathbf{A})$ :

1.  $r(\mathbf{A})$  is the largest order of those submatrices of  $\mathbf{A}$  with non null determinants.
2.  $0 \leq r(\mathbf{A}) \leq \min(n, k)$
3.  $r(\mathbf{A}) = r(\mathbf{A}')$
4.  $r(\mathbf{A}'\mathbf{A}) = r(\mathbf{A}\mathbf{A}') = r(\mathbf{A})$
5. If  $n=k$  then  $r(\mathbf{A})=k$  if and only if  $\mathbf{A}$  is non-singular

Example:

$$\mathbf{A} = \begin{pmatrix} 1 & -1 & 3 \\ -2 & 2 & 5 \\ 1 & -1 & 9 \end{pmatrix} \quad \det(\mathbf{A}) = 0 \quad \begin{vmatrix} 1 & -1 \\ -2 & 2 \end{vmatrix} = 0 \quad \begin{vmatrix} -1 & 3 \\ 2 & 5 \end{vmatrix} = -11 \neq 0$$

Thus the  $r(\mathbf{A})=2$

# Matrix algebra: eigenvalue and eigenvector

If  $\mathbf{A}$  is a **square matrix** of order  $n$ , in some problems we are interested in finding a vector  $\mathbf{x}$  and a scalar  $\lambda$  which satisfy the following property:

$$\mathbf{A}\mathbf{x} = \lambda\mathbf{x} \implies (\mathbf{A} - \lambda\mathbf{I})\mathbf{x} = \mathbf{0}$$

eigenvalue

eigenvector

A trivial solution is  $\mathbf{x}=\mathbf{0}$ , any  $\lambda \in \mathbf{R}$

# Matrix algebra: the eigenvalues

The  $n$  **eigenvalues** of  $\mathbf{A}$   $\lambda_1, \dots, \lambda_n$  are the  $n$  solutions of the characteristic equation

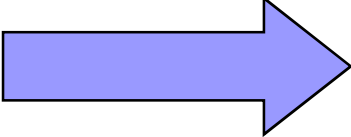
$$|\mathbf{A} - \lambda\mathbf{I}| = 0$$

Properties of the eigenvalues of  $\mathbf{A}$ :

1.  $|\mathbf{A}| = \prod_i \lambda_i$
2.  $tr(\mathbf{A}) = \sum_i \lambda_i$
3.  $r(\mathbf{A})$  equals the number of non-zero eigenvalues
4. The set of all eigenvectors for an eigenvalue  $\lambda_i$  is called the eigenspace of  $\mathbf{A}$  for  $\lambda_i$
5. Any symmetric  $n \times n$  matrix  $\mathbf{A}$  can be written as  $\mathbf{A} = \mathbf{\Gamma}\mathbf{\Lambda}\mathbf{\Gamma}' = \sum_i \lambda_i \gamma_{(i)} \gamma_{(i)}'$  where  $\mathbf{\Lambda}$  is a diagonal matrix of eigenvalues of  $\mathbf{A}$  and  $\mathbf{\Gamma}$  is an orthogonal matrix whose columns are eigenvectors with  $\gamma_{(i)}' \gamma_{(i)} = 1$

# Matrix algebra: examples

Example:

$$\mathbf{A} = \begin{pmatrix} 2 & -2 & 3 \\ 1 & 1 & 1 \\ 1 & 3 & -1 \end{pmatrix} \quad \text{The characteristic equation is:} \quad \begin{vmatrix} 2-\lambda & -2 & 3 \\ 1 & 1-\lambda & 1 \\ 1 & 3 & -1-\lambda \end{vmatrix} = 0$$


By computing the determinant we have:

$$\begin{aligned} (2-\lambda) \cdot \begin{vmatrix} 1-\lambda & 1 \\ 3 & -1-\lambda \end{vmatrix} - 1 \cdot \begin{vmatrix} -2 & 3 \\ 3 & -1-\lambda \end{vmatrix} + 1 \cdot \begin{vmatrix} -2 & 3 \\ 1-\lambda & 1 \end{vmatrix} = \\ = (2-\lambda)[-1+\lambda^2-3] - 2 - 2\lambda + 9 - 2 - 3 + 3\lambda = (\lambda+2)(\lambda-1)(3-\lambda) = 0 \end{aligned}$$

The solutions represent the 3 eigenvalues of  $\mathbf{A}$ :

$$\lambda_1 = 1 \quad \lambda_2 = -2 \quad \lambda_3 = 3$$



# Matrix algebra: the dominant eigenvalue

The eigenvalue with maximum absolute value  $\lambda_3=3$  is called dominant

There is an infinite number of eigenvectors  $\mathbf{x}$  which satisfy  $(\mathbf{A}-3\mathbf{I})\mathbf{x} = \mathbf{0}$

$$\begin{pmatrix} -1 & -2 & 3 \\ 1 & -2 & 1 \\ 1 & 3 & -4 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

## Lab using R: matrix

Please, open R and  
follow the professor's instruction