## Linear Algebra

**Tutorial** 

Inner Product, Outer Product and Norms

(1) Angle between two nonnegative vectors. Let x and y be two nonzero n-vectors with nonnegative entries, i.e., each  $x_i \geq 0$  and each  $y_i \geq 0$ . Show that the angle between x and y lies between 0 and 90°. Draw a picture for the case when n = 2, and give a short geometric explanation. When are x and y orthogonal?

- Initialize two vectors(2-D) with nonnegative entries and plot them.(use matplotlib)
- Modify the entries of vectors and observe the angle between the vectors.

(2) Distance between Boolean vectors. Suppose that x and y are Boolean n-vectors, which means that each of their entries is either 0 or 1. What is their distance ||x - y||?

- Initialize two vectors with boolean entries(0s and 1s)
- Calculate the distance between them
- Calculate hamming distance (Use scipy)

(3) Reverse triangle inequality. Suppose a and b are vectors of the same size. The triangle inequality states that  $||a+b|| \le ||a|| + ||b||$ . Show that we also have  $||a+b|| \ge ||a|| - ||b||$ . Hints. Draw a picture to get the idea. To show the inequality, apply the triangle inequality to (a+b)+(-b).

- Initialize two random vectors
- Verify triangle inequality and reverse triangular inequality

(4) When is the outer product symmetric? Let a and b be n-vectors. The inner product is symmetric, i.e., we have  $a^Tb = b^Ta$ . The outer product of the two vectors is generally not symmetric; that is, we generally have  $ab^T \neq ba^T$ . What are the conditions on a and b under which  $ab = ba^T$ ? You can assume that all the entries of a and b are nonzero. (The conclusion you come to will hold even when some entries of a or b are zero.) Hint. Show that  $ab^T = ba^T$  implies that  $a_i/b_i$  is a constant (i.e., independent of i).

## Matrices

(5) Block matrix. Assuming the matrix

$$K = \left[egin{array}{cc} I & A^T \ A & 0 \end{array}
ight]$$

makes sense, which of the following statements must be true? ('Must be true' means that it follows with no additional assumptions.)

- (a) K is square.
- (b) A is square or wide.
- (c) K is symmetric, i.e.,  $K^T = K$ .
- The identity and zero submatrices in K have the same dimensions.
- (e) The zero submatrix is square.

- Matrix sizes. Suppose A, B, and C are matrices that satisfy  $A + BB^T = C$ . Determine (6) which of the following statements are necessarily true. (There may be more than one true statement.)

  - (b) A and B have the same dimensions.
  - (c) A, B, and C have the same number of rows.

- - (d) B is a tall matrix.

- (a) A is square.

(7) Multiplication by a diagonal matrix. Suppose that A is an  $m \times n$  matrix, D is a diagonal matrix, and B = DA. Describe B in terms of A and the entries of D. You can refer to the rows or columns or entries of A.

- Create a matrix A and diagonal matrix D
- Calculate B = DA
- Observe the entries of B

**Trace of a square matrix.** The *trace* of a square matrix  $A \in \mathbb{R}^{n \times n}$ ,  $A = (a_{ij})$ , is defined to be the sum of its diagonal elements:

$$\operatorname{trace}(A) = \sum_{i=1}^{n} a_{ii}$$
.

It's obvious that trace is linear,

$$trace(A + B) = trace(A) + trace(B)$$
  $trace(\alpha A) = \alpha trace(A)$ ,

and that

(8)

$$trace(A^T) = trace(A)$$
.

Less obvious is the following fact.

a) For  $A, B \in \mathbb{R}^{n \times n}$ , show that

$$trace(AB) = trace(BA)$$
.

- Create a matrix A and B
- Calculate trace of A,B,A+B,AB and BA (use numpy.trace)
- Observe the values

b) The properties of the trace allow us to define an inner product of two square matrices A and B by

$$\langle A, B \rangle = \operatorname{trace}(AB^T).$$

This inner product then defines a norm of A as

$$||A|| = \{\operatorname{trace}(AA^T)\}^{1/2}.$$

What is ||A|| in terms of the entries of A?

c) We can define a vectorize function,  $\operatorname{vec}(A) \in \mathbb{R}^{n^2}$ , which, given the matrix A, stacks up the columns of A into a vector of length  $n^2$  (first column followed by the second column, etc.) Show that  $\langle A, B \rangle = \operatorname{trace}(AB^T) = \operatorname{vec}(A)^T \operatorname{vec}(B)$ .

- Calculate values of <A,B> and <A,A>
- Flatten the matrices A,B and calculate their dot product
- Calculate frobenius norm of matrix A and compare it with ||A||s

Eigenvalues and Eigenvectors

(9) Find the eigenvalues and eigenvectors of the matrix  $A = \begin{bmatrix} 1 & -1 \\ 2 & 4 \end{bmatrix}$ . Verify that the trace equals the sum of the eigenvalues, and the determinant equals their product.

- Initialize A
- Calculate eigenvalues and eigenvectors
- Calculate sum and product of eigenvalues and compare them to trace and determinant of A

(10) If we shift to A - 7I, what are the eigenvalues and eigenvectors and how are they related to those of A?

$$B = A - 7I = \begin{bmatrix} -6 & -1 \\ 2 & -3 \end{bmatrix}.$$

- Calculate B
- Calculate eigenvalues and eigenvectors of B
- Compare the above to eigenvalues and eigenvectors of A

- (11) Suppose that  $\lambda$  is an eigenvalue of A, and x is its eigenvector:  $Ax = \lambda x$ .
  - (a) Show that this same x is an eigenvector of B = A 7I, and find the eigenvalue.
  - (b) Assuming  $\lambda \neq 0$ , show that x is also an eigenvector of  $A^{-1}$ —and find the eigenvalue.

- Calculate inverse of A
- Calculate eigenvalues and eigenvectors for inverse of A
- Compare the above to eigenvalues and eigenvectors of A

# Least Squares

## (12) Find the least squares solution to the problem with data,

$$A = \begin{bmatrix} 2 & 0 \\ -1 & 1 \\ 0 & 2 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

- Find pseudo inverse of A by your own code and using inbuilt function
- Find the solution by multiplying pseudo inverse of A and B

# Additional Questions

Norm of matrix-vector product. Suppose A is an  $m \times n$  matrix and x is an n-vector. A famous inequality relates ||x||, ||A||, and ||Ax||:

The left-hand side is the (vector) norm of the matrix-vector product; the right-hand side is the (scalar) product of the matrix and vector norms. Show this inequality. *Hints*. Let

 $||Ax|| \leq ||A|| ||x||.$ 

is the (scalar) product of the matrix and vector norms. Show this inequality. *Hints*. Let  $a_i^T$  be the *i*th row of *A*. Use the Cauchy–Schwarz inequality to get  $(a_i^T x)^2 \leq ||a_i||^2 ||x||^2$ . Then add the resulting *m* inequalities.

- 15) Express the following statements in matrix language. You can assume that all matrices mentioned have appropriate dimensions. Here is an example: "Every column of C is a linear combination of the columns of B" can be expressed as "C = BF for some matrix F".
- There can be several answers; one is good enough for us.

  a) Suppose Z has n columns. For each i, row i of Z is a linear combination of rows  $i, \ldots, n$ 
  - of Y.
    b) W is obtained from V by permuting adjacent odd and even columns (i.e., 1 and 2, 3
  - and 4, ...).
    c) Each column of P makes an acute angle with each column of Q.
  - d) Each column of P makes an acute angle with the corresponding column of Q.
  - e) The first k columns of A are orthogonal to the remaining columns of A.