

- Functions of N squares Matrices

• Power Series

+ A square matrix can multiply itself (ie, $(N \times N)(N \times N) = (N \times N)$), so we can raise it to a power $A^n = \underbrace{AA \cdots A}_{n \text{ times}}$

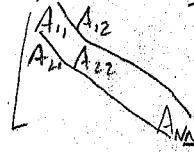
+ Many functions can be written as a Taylor series, ie, $f(x) = \sum_{n=0}^{\infty} \frac{d^n f(0)}{n!} x^n = \sum f_n x^n$. Give some examples.

+ We can define a function $f(A)$ for a square matrix A in terms of the Taylor series of $f(x)$.

$$f(A) = \sum_n f_n A^n$$

• Trace: a single number

+ The elements $A_{11}, A_{22}, \dots, A_{NN}$ lie on the main diagonal of a square matrix



+ The trace is the sum of the diagonal elements $\text{tr}(A) = A_{ii}$.

+ The trace is a linear operation: $\text{tr}(A+B) = \text{tr}A + \text{tr}B$ and $\text{tr}(cA) = c\text{tr}A$.

+ Cyclic Property: The trace is unchanged by cyclic permutations

$$\text{tr}(A_1 A_2 \cdots A_N) = \text{tr}(A_N A_1 \cdots A_{N-1}). \text{ To understand this, note that}$$

$$\text{tr}(AB) = (AB)_{ii} = A_{ij} B_{ji} = B_{ji} A_{ij} = (BA)_{jj} = \text{tr}(BA).$$

+ Can you show $\text{tr}(A^T) = \text{tr}(A)$ and $\text{tr}(A^*) = \text{tr}(A)^*$?

• Determinant: another single number

+ The determinant is given by a sum of products of one term per row, with a sign. It is defined

$$\det(A) = |A| = \sum_P (-1)^{\text{sign}(P)} A_{i_1 j_1} A_{i_2 j_2} \cdots A_{i_N j_N} \approx \sum \begin{bmatrix} \cdot & \cdot & \cdot & \cdot \\ \cdot & \circ & \cdot & \cdot \\ \cdot & \cdot & \circ & \cdot \\ \cdot & \cdot & \cdot & \cdot \end{bmatrix}$$

where P is a permutation $\{j_1, j_2, \dots, j_N\}$ of $\{1, 2, 3, \dots, N\}$. $\text{Sign}(P) = +1$ if $\{j_1, j_2, \dots, j_N\} = \{1, 2, \dots, N\}$ and changes sign each time a value is permuted.

+ Smaller matrices are fairly simple:

$$2 \times 2: \det A = A_{11}A_{22} - A_{12}A_{21} \quad \text{Graphically, } \det = \begin{vmatrix} * & * \\ * & * \end{vmatrix}$$

is a good mnemonic

$$3 \times 3: \det A = A_{11}A_{22}A_{33} + A_{12}A_{23}A_{31} + A_{13}A_{21}A_{32} - A_{11}A_{23}A_{32} - A_{12}A_{13}A_{33} - A_{13}A_{21}A_{31}$$

$$\det = \begin{vmatrix} * & * & * \\ * & * & * \\ * & * & * \end{vmatrix}$$

You can see that larger one will rapidly get complicated.

+ The way to calculate larger ones is to notice that we can write (reorganizing the sum)

$$\det(A) = A_{11}C_{11} + A_{12}C_{12} + \dots + A_{1N}C_{1N}$$

where each of the factors C_{ij} looks like a smaller determinant

$$C_{11} = \sum (\pm) A_{2j_2}A_{3j_3} \dots A_{Nj_N} \quad \text{for example}$$

+ In fact, the cofactor C_{ij} of the ij -element is

$(-1)^{i+j}$ times the determinant of the $(N-1) \times (N-1)$ matrix formed by removing the i^{th} row and j^{th} column of A :

$$C_{ij} = (-1)^{i+j} \det \begin{bmatrix} \cdots & \cdots & \cdots \\ \hline \cdots & \cancel{i} & \cdots \\ \cdots & \cdots & \cdots \end{bmatrix}$$

+ Then $\det(A)$ is given by summing the element \times the cofactor either across any row or down any column:

$$\det(A) = A_{11}C_{11} + A_{12}C_{12} + \dots + A_{1N}C_{1N}$$

$$= A_{21}C_{21} + A_{22}C_{22} + \dots + A_{2N}C_{2N} = \dots$$

$$= A_{11}C_{11} + A_{21}C_{11} + \dots + A_{N1}C_{11} = \dots$$

This is still difficult usually, and computers help a lot.

+ There are a few simple properties of the determinant:

$$\det(A^T) = \det(A), \det(A^*) = \det(A^T) = \det(A)^*, \det(\lambda A) = \lambda^N \det(A),$$

the determinant changes sign if you interchange 2 rows or 2 columns

- + These properties (unproved here) can help evaluating determinants:
 - a) if 2 rows (or 2 columns) are multiples of each other, $\det(A) = 0$.
 - b) if you add a multiple of one row (or column) to another, the determinant is unchanged
 - c) $\det(AB) = \det(A)\det(B)$
- + One strategy for a large determinant is to add multiples of other rows (or columns) to a particular row (or column) in order to reduce as many elements as possible to zero. That simplifies the sum over cofactors
- + Work some examples.