

12. Kronecker Product and the Vec Operator

The Kronecker Product

We use partitioned matrices to prove a few results involving the Kronecker product and the Vec operator.

Suppose $\mathbf{A} = (a_{ij})_{m \times n}$ and \mathbf{B} is a $p \times q$ matrix. Then the Kronecker product \otimes is defined to be the $mp \times nq$ matrix

$$\mathbf{A} \otimes \mathbf{B} = \begin{bmatrix} a_{11}\mathbf{B} & a_{12}\mathbf{B} & \cdots & a_{1n}\mathbf{B} \\ a_{21}\mathbf{B} & a_{22}\mathbf{B} & \cdots & a_{2n}\mathbf{B} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1}\mathbf{B} & a_{m2}\mathbf{B} & \cdots & a_{mn}\mathbf{B} \end{bmatrix}$$

Example If $\mathbf{A} = \begin{bmatrix} 1 & 2 & 0 \\ 3 & 0 & 1 \end{bmatrix}$ and $\mathbf{B} = \begin{bmatrix} 5 & 2 \\ 1 & 3 \end{bmatrix}$, then $\mathbf{A} \otimes \mathbf{B} = \begin{bmatrix} 5 & 2 & 10 & 4 & 0 & 0 \\ 1 & 3 & 2 & 6 & 0 & 0 \\ 15 & 6 & 0 & 0 & 5 & 2 \\ 3 & 9 & 0 & 0 & 1 & 3 \end{bmatrix}$

If $\mathbf{b}' = [2 \ 3 \ 1]$, then $\mathbf{I}_2 \otimes \mathbf{b}' = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \otimes [2 \ 3 \ 1] = \begin{bmatrix} 2 & 3 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 3 & 1 \end{bmatrix}$

Properties

1. In general $\mathbf{A} \otimes \mathbf{B} \neq \mathbf{B} \otimes \mathbf{A}$; the two products will in general not even have the same dimensions.
2. $(\mathbf{A} \otimes \mathbf{B})' = \mathbf{A}' \otimes \mathbf{B}'$

Proof

$$(\mathbf{A} \otimes \mathbf{B})' = \begin{bmatrix} a_{11}\mathbf{B} & a_{12}\mathbf{B} & \cdots & a_{1n}\mathbf{B} \\ a_{21}\mathbf{B} & a_{22}\mathbf{B} & \cdots & a_{2n}\mathbf{B} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1}\mathbf{B} & a_{m2}\mathbf{B} & \cdots & a_{mn}\mathbf{B} \end{bmatrix}^T = \begin{bmatrix} a_{11}\mathbf{B}' & a_{21}\mathbf{B}' & \cdots & a_{m1}\mathbf{B}' \\ a_{12}\mathbf{B}' & a_{22}\mathbf{B}' & \cdots & a_{m2}\mathbf{B}' \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n}\mathbf{B}' & a_{2n}\mathbf{B}' & \cdots & a_{mn}\mathbf{B}' \end{bmatrix} = \mathbf{A}' \otimes \mathbf{B}'$$

3. $\mathbf{A} \otimes (\mathbf{B} + \mathbf{C}) = (\mathbf{A} \otimes \mathbf{B}) + (\mathbf{A} \otimes \mathbf{C})$

Proof: Left as an exercise.

4. $(\mathbf{A} \otimes \mathbf{B})(\mathbf{C} \otimes \mathbf{D}) = \mathbf{AC} \otimes \mathbf{BD}$
 $m \times n$ $p \times q$ $n \times r$ $q \times s$

Exercise: check that the dimensions of the matrices are such that all products involved exist.

Proof:

$$(\mathbf{A} \otimes \mathbf{B})(\mathbf{C} \otimes \mathbf{D}) = \begin{bmatrix} a_{11}\mathbf{B} & a_{12}\mathbf{B} & \cdots & a_{1n}\mathbf{B} \\ a_{21}\mathbf{B} & a_{22}\mathbf{B} & \cdots & a_{2n}\mathbf{B} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1}\mathbf{B} & a_{m2}\mathbf{B} & \cdots & a_{mn}\mathbf{B} \end{bmatrix} \begin{bmatrix} c_{11}\mathbf{D} & c_{12}\mathbf{D} & \cdots & c_{1r}\mathbf{D} \\ c_{21}\mathbf{D} & c_{22}\mathbf{D} & \cdots & c_{2r}\mathbf{D} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1}\mathbf{D} & c_{n2}\mathbf{D} & \cdots & c_{nr}\mathbf{D} \end{bmatrix}$$

The (i, j) th block of this product is $\sum_{k=1}^n a_{ik}c_{kj}\mathbf{BD} = [\mathbf{AC}]_{ij}\mathbf{BD}$, which gives the required result.

5. $(\mathbf{A} \otimes \mathbf{B})^{-1} = \mathbf{A}^{-1} \otimes \mathbf{B}^{-1}$ where \mathbf{A} and \mathbf{B} are $n \times n$ and $m \times m$ invertible matrices.

Proof: From (4), we have $(\mathbf{A}^{-1} \otimes \mathbf{B}^{-1})(\mathbf{A} \otimes \mathbf{B}) = \mathbf{I}_n \otimes \mathbf{I}_m = \mathbf{I}_{nm}$.

6. $tr(\mathbf{A} \otimes \mathbf{B}) = tr(\mathbf{A})tr(\mathbf{B})$ where \mathbf{A} and \mathbf{B} are square matrices.

Proof: Left as an exercise.

7. For the square matrices $\mathbf{A}_{(m \times m)}$ and $\mathbf{B}_{(n \times n)}$, we have $|\mathbf{A} \otimes \mathbf{B}| = |\mathbf{A}|^n |\mathbf{B}|^m$.

Proof: Omitted.

Remark: We can use eigenvalues to prove this result. Specifically, if λ_A and λ_B are eigenvalues of \mathbf{A} and \mathbf{B} respectively, then $\lambda_A \lambda_B$ is an eigenvalue of $\mathbf{A} \otimes \mathbf{B}$. If λ_A and λ_B are eigenvalues of \mathbf{A} and \mathbf{B} respectively, with associated eigenvectors \mathbf{a} and \mathbf{b} , then we have $\mathbf{Aa} = \lambda_A \mathbf{a}$ and $\mathbf{Bb} = \lambda_B \mathbf{b}$. This means that

$$\begin{aligned} (\mathbf{A} \otimes \mathbf{B})(\mathbf{a} \otimes \mathbf{b}) &= (\mathbf{Aa} \otimes \mathbf{Bb}) \\ &= (\lambda_A \mathbf{a} \otimes \lambda_B \mathbf{b}) \\ &= \lambda_A \lambda_B (\mathbf{a} \otimes \mathbf{b}) \end{aligned}$$

which says that $\lambda_A \lambda_B$ is an eigenvalue of $\mathbf{A} \otimes \mathbf{B}$. Furthermore, the determinant of a square matrix is the product of its eigenvalues. Applying this to the m eigenvalues of \mathbf{A} and the n eigenvalues of \mathbf{B} completes the proof.

The vec operator

The vec operator is used to ‘vectorize’ matrices. For any $m \times n$ matrix $\mathbf{A} = (a_{ij})_{m \times n} = [\mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_n]$, where \mathbf{a}_i is the i th column of \mathbf{A} , $i = 1, 2, \dots, n$, define $vec(\mathbf{A})$ to be the $mn \times 1$ vector

$$vec(\mathbf{A}) = \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \vdots \\ \mathbf{a}_n \end{bmatrix}.$$

Example: if $\mathbf{A} = \begin{bmatrix} 1 & 2 & 0 \\ 3 & 0 & 1 \end{bmatrix}$, then $vec(\mathbf{A}) = \begin{bmatrix} 1 \\ 3 \\ 2 \\ 0 \\ 0 \\ 1 \end{bmatrix}$.

Properties

1. $vec(\mathbf{A} + \mathbf{B}) = vec(\mathbf{A}) + vec(\mathbf{B})$

Proof: Left as an exercise.

2. $vec(\mathbf{A} \mathbf{B}) = (\mathbf{I}_p \otimes \mathbf{A})vec(\mathbf{B}) = (\mathbf{B}' \otimes \mathbf{I}_m)vec(\mathbf{A})$

Proof: The columns of the product \mathbf{AB} can be written as $\mathbf{AB} = \mathbf{A}[\mathbf{b}_1 \ \mathbf{b}_2 \ \cdots \ \mathbf{b}_p]$. Therefore,

$$vec(\mathbf{AB}) = \begin{bmatrix} \mathbf{Ab}_1 \\ \mathbf{Ab}_2 \\ \vdots \\ \mathbf{Ab}_p \end{bmatrix} = \begin{bmatrix} \mathbf{A} & \mathbf{0} & \cdots & \mathbf{0} \\ \mathbf{0} & \mathbf{A} & \cdots & \mathbf{0} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{0} & \mathbf{0} & \cdots & \mathbf{A} \end{bmatrix} \begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \\ \vdots \\ \mathbf{b}_p \end{bmatrix} = (\mathbf{I}_p \otimes \mathbf{A})vec(\mathbf{B}).$$

Since $\mathbf{A}\mathbf{b}_i = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_n] \begin{bmatrix} b_{1i} \\ b_{2i} \\ \vdots \\ b_{ni} \end{bmatrix} = b_{1i}\mathbf{a}_1 + b_{2i}\mathbf{a}_2 + \cdots + b_{ni}\mathbf{a}_n$, we have

$$\text{vec}(\mathbf{AB}) = \begin{bmatrix} \mathbf{A}\mathbf{b}_1 \\ \mathbf{A}\mathbf{b}_2 \\ \vdots \\ \mathbf{A}\mathbf{b}_p \end{bmatrix} = \begin{bmatrix} b_{11}\mathbf{I}_m & b_{21}\mathbf{I}_m & \cdots & b_{n1}\mathbf{I}_m \\ b_{12}\mathbf{I}_m & b_{22}\mathbf{I}_m & \cdots & b_{n2}\mathbf{I}_m \\ \vdots & \vdots & \ddots & \vdots \\ b_{1p}\mathbf{I}_m & b_{2p}\mathbf{I}_m & \cdots & b_{np}\mathbf{I}_m \end{bmatrix} \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \vdots \\ \mathbf{a}_n \end{bmatrix} = (\mathbf{B}' \otimes \mathbf{I}_m) \text{vec}(\mathbf{A}).$$

3. $\text{tr}(\mathbf{A}\mathbf{B}) = \text{vec}(\mathbf{A}')' \text{vec}(\mathbf{B})$
 $m \times n \quad n \times m$

Proof: Let $\tilde{\mathbf{a}}_i$ denote the i th column of \mathbf{A}' (i.e., the transpose of the i th row of \mathbf{A}).

Then

$$\text{tr}[\mathbf{AB}] = \sum_{i=1}^m \tilde{\mathbf{a}}_i' \mathbf{b}_i = [\tilde{\mathbf{a}}_1' \quad \tilde{\mathbf{a}}_2' \quad \cdots \quad \tilde{\mathbf{a}}_m'] \begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \\ \vdots \\ \mathbf{b}_m \end{bmatrix} = \text{vec}(\mathbf{A}')' \text{vec}(\mathbf{B}).$$