## Matrix Algebra 3 (Loose ends)

:Trace
Definition: Let $A \in \mathbb{R}^{m \times m}$. The trace of $A$ is the sum of its diagonal components, i.e.

$$
\operatorname{tr}(A) \equiv \sum_{i=1}^{m} a_{i i}
$$

Properties:

1. $(\forall c \in \mathbb{R}) \operatorname{tr}(c A)=c \cdot \operatorname{tr}(A) ; \operatorname{tr}(A+B)=\operatorname{tr}(A)+\operatorname{tr}(B)$
2. $\operatorname{tr}(A)=\operatorname{tr}\left(A^{\prime}\right)$
3. if both products exist, $\operatorname{tr}(A B)=\operatorname{tr}(B A)$
4. if $A$ is idempotent, $\operatorname{tr}(A)=r k(A)$
:Partitioned Matrices
We can partition a matrix into submatrices. For example, if

$$
A=\left[\begin{array}{cccc}
4 & 0 & 2 & -1 \\
6 & 5 & 1 & 1 \\
-3 & 2 & 0 & 5
\end{array}\right]=\left[\begin{array}{ll}
A_{11} & A_{12} \\
A_{21} & A_{22}
\end{array}\right]
$$

where

$$
\begin{aligned}
& A_{11}=\left[\begin{array}{lll}
4 & 0 & 2 \\
6 & 5 & 1
\end{array}\right] A_{12}=\left[\begin{array}{c}
-1 \\
1
\end{array}\right] \\
& A_{21}=\left[\begin{array}{lll}
-3 & 2 & 0
\end{array}\right] A_{22}=[5]
\end{aligned}
$$

We require that the submatrices appearing in any given row (column) must themselves have all the same number of rows (columns).

The usual rules for the addition and multplication of matrices apply directly to submatrices provided the operations are always applied to conformable submatrices.

- If $A$ and $B$ are two partitioned matrices, then

$$
A+B=\left[\begin{array}{ll}
A_{11}+B_{11} & A_{12}+B_{12} \\
A_{21}+B_{21} & A_{22}+B_{22}
\end{array}\right]
$$

provided that all sums $A_{i j}+B_{i j}$ exist.

- Similarly,

$$
\begin{aligned}
A B & =\left[\begin{array}{ll}
A_{11} & A_{12} \\
A_{21} & A_{22} \\
A_{31} & A_{32}
\end{array}\right]\left[\begin{array}{ll}
B_{11} & B_{12} \\
B_{21} & B_{22}
\end{array}\right] \\
& =\left[\begin{array}{ll}
A_{11} B_{11}+A_{12} B_{21} & A_{11} B_{12}+A_{12} B_{22} \\
A_{21} B_{11}+A_{22} B_{21} & A_{21} B_{12}+A_{22} B_{22} \\
A_{31} B_{11}+A_{32} B_{21} & A_{31} B_{12}+A_{32} B_{22}
\end{array}\right]
\end{aligned}
$$

- In particular, if we write $X=\left[\begin{array}{ll}X_{1} & X_{2}\end{array}\right]$ where $X_{i} \in \mathbb{R}^{n x K_{i}}$, then

$$
X^{\prime} X=\left[\begin{array}{ll}
X_{1}^{\prime} X_{1} & X_{1}^{\prime} X_{2} \\
X_{2}^{\prime} X_{1} & X_{2}^{\prime} X_{1}
\end{array}\right]
$$

and

$$
\left(X^{\prime} X\right)^{-1} X^{\prime} X_{1}=\left[\begin{array}{c}
I_{K_{1}} \\
0
\end{array}\right]
$$

## :Quadratic Forms

Definition: Suppose $x \in \mathbb{R}^{m}$ and $A \in \mathbb{R}^{m x m}$ with $A$ symmetric. Then $x^{\prime} A x$ is called a quadratic form.
Rks:

- $x^{\prime} A x \in \mathbb{R}$
- It is not restrictive to assume that $A$ is symmetric. Using $x^{\prime} A x=x^{\prime} A^{\prime} x=x^{\prime} \tilde{A} x$ where $\widetilde{A}=\left(A+A^{\prime}\right) / 2$, we can always replace $A$ with $\widetilde{A}$.
- For $m=2$,

$$
x^{\prime} A x=a_{11} x_{1}^{2}+2 a_{12} x_{1} x_{2}+a_{22} x_{2}^{2}
$$

Definition: The matrix $A$ is called positive definite (p.d.) if $x^{\prime} A x>0$ for $\forall x \neq 0$.
Definition: The matrix $A$ is called nonnegative definite (n.n.d) if $x^{\prime} A x \geq 0$ for $\forall x$.

Definition: The matrix $A$ is called positive semidefinite (p.s.d.) if $x^{\prime} A x \geq 0$ for $\forall x$ and for some $x \neq 0$ we have $x^{\prime} A x=0$.
Rks:

- By changing the sign on the inequalities, we can define the set of negative definite, nonpositive definite, and negative semidefinite matrices.
- Many (most?) authors do not distinguish between nonnegative definite and positive semidefinite (or nonpositive definite and negative semidefinite matrices).

Theorems:

1. $A$ is positive definite iff we can write $A=R R^{\prime}$ for some nonsingular matrix $R$.
2. $A$ is nonnegative definite iff we can write $A=R R^{\prime}$ for some matrix $R$.
3. If $A$ is positive definite, and $B$ has full column rank, then $B^{\prime} A B$ is positive definite.
Definition: If we can write $A=R R^{\prime}$ for some matrix $R$, then $R$ is called a square-root matrix of $A$.
Rks:

- If $A$ has a square-root, then $x^{\prime} A x=x^{\prime} R R^{\prime} x=c^{\prime} c$ where $c \equiv R^{\prime} x$.
- There are lots ( $2 \cdot m!$ ) square root matrices. If $R$ is a square root, then so is $\widetilde{R}=R \Theta$ for any orthogonal matrix $\Theta$ (i.e. $\left.\Theta^{-1}=\Theta^{\prime}\right)$.

Definition: If $A, B \in R^{m \times m}$, we say " $A$ is bigger than $B$ (in the matrix sense)", written $A \geq B$, if $A-B$ is n.n.d.
Rk : The matrix $\geq$ is a partial ordering.
Theorem: If $A, B$ are invertible, then $A \geq B$ iff $B^{-1} \geq A^{-1}$.
Definition: For any $c \in \mathbb{R},\left\{x: x^{\prime} A x=c\right\}$ is called a level curve.
Rks:

- For $m=2$, if $A$ is p.d., the level curves form ellipses.
- For $m>2$, if $A$ is p.d., the level curves form ellipsoids.


## :Spectral Representation

Theorem: Every symmetric matrix $A \in \mathbb{R}^{m x m}$ can be expressed as

$$
A=P \Lambda P^{\prime}
$$

where $\Lambda \in \mathbb{R}^{m \times m}$ is a diagonal matrix, i.e.

$$
\Lambda=\left(\begin{array}{cccc}
\lambda_{1} & 0 & \cdots & 0 \\
0 & \lambda_{2} & \cdots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
0 & 0 & \cdots & \lambda_{m}
\end{array}\right) \equiv \operatorname{diag}\left(\lambda_{i}\right)
$$

and $P \in \mathbb{R}^{m \times m}$ is an orthogonal matrix (i.e. $P P^{\prime}=I$ ).
Rks:

- The $\lambda_{i}$ are called eigenvalues and the corresponding columns $P_{i}$ are called eigenvectors.


## Properties:

1. $A$ is p.d. (n.n.d.) iff each of its eigenvalues are positive (nonnegative)
2. A choice for the square-root matrix of $A$ is given by $A^{1 / 2} \equiv P \operatorname{diag}\left(\lambda_{i}^{1 / 2}\right) P^{\prime}$.
3. $A^{m}=P \operatorname{diag}\left(\lambda_{i}^{m}\right) P^{\prime}$ for all integers $m$.
4. If $A$ is symmetric and idempotent, then all its eigenvalues are either 0 or 1.
