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## ANIMAL BREEDING NOTES

## CHAPTER 4

 DEFINITE, ORTHOGONAL, AND IDEMPOTENT MATRICES
## Definitions

Definite matrices are defined for symmetric matrices only. Let $A$ be an $n \times n$ symmetric matrix and $x^{\prime} A x$ be a quadratic form. Then, the symmetric matrix $A$ and the quadratic form $x^{\prime} A x$ are said to be:
a) positive definite (p.d.),
if $x^{\prime} A x>0$ for all $x \neq 0$,
b) positive semi-definite (p.s.d.),
if $x^{\prime} A x \geq 0$ for all $x \neq 0$, with $x^{\prime} A x=0$ for at least one $x \neq 0$,
c) non-negative definite (n.n.d),
if $x^{\prime} A x \geq 0$ for all $x \neq 0$,
d) negative definite (n.d.),
if $x^{\prime} A x<0$ for all $x \neq 0$,
e) negative semi-definite (n.s.d.),
if $x^{\prime} A x \leq 0$ for all $x \neq 0$, with $x^{\prime} A x=0$ for at least one $x \neq 0$, and
f) non-positive definite (n.p.d.),
if $x^{\prime} A x \leq 0$ for all $x \neq 0$.

## Properties of positive definite (p.d.) matrices

(1) A symmetric matrix A is p.d. if and only if all the characteristic roots of $A$ are positive.

Proof: (by contradiction)
$\Leftarrow\left\{\lambda_{\mathrm{i}}>0\right\} \Rightarrow$ A p.d.

Let P be an orthogonal matrix that diagonalizes A , i.e.,

$$
\mathrm{P}^{\prime} \mathrm{AP}=\mathrm{D}=\operatorname{diag}\left\{\lambda_{\mathrm{i}}\right\},
$$

where $\left\{\lambda_{i}\right\}$ are the latent roots of $A$.

Let $y=P^{\prime} x \Rightarrow x=\left(P^{\prime}\right)^{-1} y=P y$

Thus $\quad \mathrm{x}^{\prime} \mathrm{Ax}=\mathrm{y}^{\prime} \mathrm{P}^{\prime} \mathrm{APy}=\mathrm{y}^{\prime} \mathrm{Dy}=\sum_{i=1}^{n} \lambda_{\mathrm{i}} \mathrm{y}_{\mathrm{i}}{ }^{2}$
If all $\lambda_{i}>0$, then $x^{\prime} A x=y^{\prime} D y \geq 0$ for all $y$, with equality only when $y=0$, i.e., when $x=P y=P 0=$
$0 \Rightarrow \mathrm{~A}$ is p.d.
$\Rightarrow$ A p.d. $\Rightarrow\left\{\lambda_{i}>0\right\}$

Assume a characteristic root of A, e.g. $\lambda_{1}$, is not positive.
Let $y^{*}$ be the $n \times 1$ vector with the first element equal to 1 and the rest zeroes, and let $x^{*}=P y^{*}$, then
$\mathrm{x}^{*} \neq 0 \Rightarrow$ because $\mathrm{y}^{*} \neq 0$ (see 4.28, pg. 23, Goldberger, 1964).

Then,

$$
\begin{equation*}
x^{* \prime} A^{* *}=y^{*} P^{\prime} A P P y^{*}=y^{* *} D^{* *}=\sum_{i=1}^{n} \lambda_{i} y_{i}^{* 2}=\lambda_{1} \leq 0 \tag{4-3}
\end{equation*}
$$

which contradicts the assumption that A is p.d. $\Rightarrow \lambda_{1}>0$ and by induction $\Rightarrow\left\{\lambda_{i}>0\right\}$.
(2) If $A_{n \times n}$ is p.d., then
(a) $|\mathrm{A}|>0$,
(b) $\operatorname{rank}(\mathrm{A})=\mathrm{n}$, and
(c) A is non-singular.

Proof:
(a) $|\mathrm{A}|=\left|\mathrm{P}^{\prime} \mathrm{AP}\right|=|\mathrm{D}|=\lambda_{1} \lambda_{2} \ldots \quad \lambda_{\mathrm{n}}$, where $\left\{\lambda_{\mathrm{i}}>0\right\}$ by property (1) of p.d. matrices, thus,
$|\mathrm{D}|>0 \Rightarrow|\mathrm{~A}|>0$,
(b) $\operatorname{rank}(\mathrm{A})=\operatorname{rank}\left(\mathrm{P}^{\prime} \mathrm{AP}\right)$,
$=\operatorname{rank}(\mathrm{D})$,
$=n$ because $\lambda_{i}>0, i=1, \ldots, n$,
(c) A is nonsingular because $|\mathrm{A}|>0$ as proven in (a).
(3) If $A_{n \times n}$ is p.d. and $P$ is an $n \times m$ matrix with rank $(P)=m$, then $P^{\prime} A P$ is p.d.

Proof: $P^{\prime} A P$ is an $m \times m$ symmetric matrix. Consider $y_{n \times 1}, y \neq 0$, then $y^{\prime}\left(P^{\prime} A P\right) y=x^{\prime} A x$ for $x=$ Py. Because $A$ is $p . d$. and $x \neq 0$, then $x^{\prime} A x>0$. But $y^{\prime}\left(P^{\prime} A P\right) y=x^{\prime} A x$, thus $y^{\prime}\left(P^{\prime} A P\right) y>0$ for all $y \neq 0$, so, by definition, $P^{\prime} A P$ is p.d.

## Specializations of property (3)

(3.1) If $A$ is p.d. and $P$ is nonsingular, then $P^{\prime} A P$ is p.d.

Proof: same as for property (3) above.
(3.2) If $A$ is p.d., then $A^{-1}$ is p.d.

Proof: Let

$$
\begin{aligned}
& \mathrm{P}=\left(\mathrm{A}^{-1}\right)^{\prime} \mathrm{AA}^{-1} \\
&=\left(\mathrm{A}^{-1}\right)^{\prime} \\
&=\mathrm{A}^{-1} \text { because } \mathrm{A} \text { is symmetric } \\
& \Rightarrow \quad \mathrm{A}^{-1} \text { is p.d. }
\end{aligned}
$$

(3.3) If P is an $\mathrm{n} \times \mathrm{m}$ matrix with $\operatorname{rank}(\mathrm{P})=\mathrm{m}$, then $\mathrm{P}^{\prime} \mathrm{P}$ is $\mathrm{p} . \mathrm{d}$.

Proof: Consider A = I in (3) above. The identity matrix I is p.d. because

$$
x^{\prime} I x=\sum_{i=1}^{n} x_{i}^{2}>0 \quad \text { for all } x \neq 0
$$

So, we have:

$$
\mathrm{P}^{\prime} \mathrm{AP}=\mathrm{P}^{\prime} \mathrm{IP}=\mathrm{P}^{\prime} \mathrm{P} \Rightarrow \mathrm{P}^{\prime} \mathrm{P} \text { is p.d., by property }(3) \text { above. }
$$

(4) A principal submatrix of a square matrix $A$ is a submatrix whose diagonal elements coincide with the diagonals of A. A principal submatrix is obtained by deleting the appropriate rows and columns of A. If A is p.d., then every principal submatrix of A is p.d.

Proof: Without loss of generality, let B be the principal submatrix of A obtained by deleting the last $\mathrm{n}-\mathrm{m}$ rows and columns of A . Then,

$$
\mathrm{B}=\left[\begin{array}{ll}
\mathrm{I}_{\mathrm{m}} & 0_{\mathrm{m}, \mathrm{n}-\mathrm{m}}
\end{array}\right]\left[\begin{array}{cc}
\mathrm{A}_{11} & \mathrm{~A}_{12} \\
\mathrm{~A}_{12}, & \mathrm{~A}_{22}
\end{array}\right]\left[\begin{array}{c}
\mathrm{I}_{\mathrm{m}} \\
0_{\mathrm{n}-\mathrm{m}, \mathrm{~m}}
\end{array}\right]
$$

Because $\left[\begin{array}{c}I_{m} \\ 0_{n-m, m}\end{array}\right]$ is an $n \times m$ matrix of rank equal to $m$, it qualifies as the $P$ of property (3) above.
Thus, by property (3), B is p.d.
(5) A principal minor is the determinant of a principal submatrix. Then, if A is p.d., then every
principal minor of A is positive.
Proof: Let $|B|$, where B comes from (4) above, be a principal minor. Since B is p.d. by property (4), $|\mathrm{B}|>0$ by property (2).

A particular case of (5) is:
If A is p.d., then
(a) $\mathrm{a}_{\mathrm{ii}}>0$, and
(b) $\mathrm{a}_{\mathrm{i} i} \mathrm{a}_{\mathrm{ij}}-\mathrm{a}_{\mathrm{ij}}^{2}>0$ for all i and j .

Proof:
(a) Without loss of generality choose $\mathrm{B}_{\mathrm{n} \times 1}$ with a 1 in the first element and zeroes elsewhere.

Hence, $\operatorname{rank}(B)=1$. Thus, by property (4) $B^{\prime} A B=\left[a_{11}\right]$ is p.d., and by property (2) its determinant is positive, i.e.,

$$
\left|\mathrm{B}^{\prime} \mathrm{AB}\right|=\left|\mathrm{a}_{11}\right|=\mathrm{a}_{11}>0
$$

(b) Without loss of generality choose $\mathrm{B}_{\mathrm{n} \times 2}$ with 1 's in positions $(1,1)$ and $(2,2)$, and zeroes elsewhere. Hence, $\operatorname{rank}(B)=2$.

By property (4),

$$
B^{\prime} A B=\left[\begin{array}{ll}
a_{11} & a_{12} \\
a_{12} & a_{22}
\end{array}\right] \text { is p.d. }
$$

By property (2),

$$
\left|B^{\prime} A B\right|=\left|\begin{array}{ll}
a_{11} & a_{12} \\
a_{12} & a_{22}
\end{array}\right|=a_{11} a_{22}-a_{12}^{2}>0
$$

(6) If A is p.d., there exists a nonsingular matrix $P$ such that $P A P^{\prime}=I$ and $P^{\prime} P=A^{-1}$.

Proof: Let E be the orthogonal matrix such that

$$
\mathrm{E}^{\prime} \mathrm{AE}=\mathrm{D}=\operatorname{diag}\left\{\lambda_{i}\right\}
$$

and let

$$
\mathrm{T}=\operatorname{diag}\left\{\frac{1}{\sqrt{\lambda_{\mathrm{i}}}}\right\} .
$$

Define:

$$
\mathrm{P}=\mathrm{T}^{\prime} \mathrm{E}^{\prime} \text {, where } \mathrm{P} \text { is nonsingular because it is the product of nonsingular matrices. }
$$

Thus,

$$
\begin{aligned}
& \mathrm{PAP}^{\prime}=\mathrm{T}^{\prime} \mathrm{E}^{\prime} \mathrm{AET} \\
& \mathrm{PAP}^{\prime}=\mathrm{T}^{\prime} \mathrm{DT} \\
& \mathrm{PAP}^{\prime}=\operatorname{diag}\left\{\frac{1}{\sqrt{\lambda_{\mathrm{i}}}}\right\} \operatorname{diag}\left\{\lambda_{\mathrm{i}}\right\} \operatorname{diag}\left\{\frac{1}{\sqrt{\lambda_{\mathrm{i}}}}\right\} \\
& \mathrm{PAP}^{\prime}=\mathrm{I}
\end{aligned}
$$

Furthermore, from $\mathrm{PAP}^{\prime}=\mathrm{I}$ we get:

$$
\begin{aligned}
\mathrm{PAP}^{\prime} & =\mathrm{I} \\
\mathrm{P}^{\prime}\left(\mathrm{PAP}^{\prime}\right) \mathrm{P} & =\mathrm{P}^{\prime} \mathrm{IP} \\
\mathrm{P}^{\prime} \mathrm{PAP}^{\prime} \mathrm{P} & =\mathrm{P}^{\prime} \mathrm{P}
\end{aligned}
$$

Because $P$ is nonsingular, $\mathrm{P}^{\prime} \mathrm{P}$ is also nonsingular, hence $\left(\mathrm{P}^{\prime} \mathrm{P}\right)^{-1}$ exists. Thus,

$$
\begin{aligned}
\left(\mathrm{P}^{\prime} \mathrm{P}\right)^{-1} \mathrm{P}^{\prime} \mathrm{PAP}^{\prime} \mathrm{P} & =\left(\mathrm{P}^{\prime} \mathrm{P}\right)^{-1} \mathrm{P}^{\prime} \mathrm{P} \\
\mathrm{AP}^{\prime} \mathrm{P} & =\mathrm{I} \\
\mathrm{~A}^{-1} \mathrm{AP}^{\prime} \mathrm{P}= & \mathrm{A}^{-1} \mathrm{I} \\
\mathrm{P}^{\prime} \mathrm{P} & =\mathrm{A}^{-1}
\end{aligned}
$$

(7) If $A$ is p.d. of order $n$, there is a full rank $n \times n$ matrix $L$ such that $A=L L^{\prime}$.

Proof: $\mathrm{PAP}^{\prime}=\mathrm{D}$ for P orthogonal, where $\mathrm{D}=$ diagonal of order n whose elements are the eigenvalues of A (and D ). Because P is orthogonal, $\mathrm{P}^{\prime} \mathrm{P}=\mathrm{PP}^{\prime}=\mathrm{I}$. Thus,

$$
\mathrm{P}^{\prime} \mathrm{PAP}^{\prime} \mathrm{P}=\mathrm{P}^{\prime} \mathrm{DP}
$$

But since $A$ is p.d. the elements of $D=\operatorname{diag}\left\{\lambda_{i}\right\}$ are all positive, thus

$$
\begin{aligned}
& A=P^{\prime} D P \\
& A=\left(P^{\prime} D^{1 / 2}\right)\left(D^{1 / 2} P\right) \\
& A=L^{\prime}, \text { where } L^{\prime}=D^{1 / 2} P
\end{aligned}
$$

Also, note that

$$
\begin{aligned}
\mathrm{L}^{\prime} \mathrm{L} & =\mathrm{D}^{1 / 2} \mathrm{PP}^{\prime} \mathrm{D}^{1 / 2} \\
& =\mathrm{D}
\end{aligned}
$$

(8) A symmetric matrix is p.d. if and only if it can be written as $\mathrm{P}^{\prime} \mathrm{P}$ for a nonsingular P .

Proof:
(a) Necessary condition: existence of P .

Because A is symmetric, there is an orthogonal matrix Q such that

$$
\begin{aligned}
\mathrm{QAQ}^{\prime} & =\mathrm{D}=\operatorname{diag}\left\{\lambda_{i}\right\} \\
\mathrm{QAQ}^{\prime} & =\mathrm{D}^{1 / 2} \mathrm{ID}^{1 / 2} \\
\Rightarrow \quad \mathrm{D}^{-1 / 2} \mathrm{QAQ}^{\prime} \mathrm{D}^{-1 / 2} & =\mathrm{D}^{-1 / 2} \mathrm{D}^{1 / 2} \mathrm{ID}^{1 / 2} \mathrm{D}^{-1 / 2} \\
\mathrm{TAT}^{\prime} & =\mathrm{I} \quad \text { for } \mathrm{T}=\mathrm{D}^{-1 / 2} \mathrm{Q}
\end{aligned}
$$

Note: T is nonsingular because $\mathrm{D}^{-1 / 2}$ and Q are, which implies that $\left(\mathrm{D}^{-1 / 2}\right)^{-1}$ and $\mathrm{Q}^{-1}$ exist. If T is nonsingular, $\mathrm{T}^{-1}=\mathrm{Q}^{-1} \mathrm{D}^{1 / 2}$ exists, because $\mathrm{Q}^{-1}$ and $\left(\mathrm{D}^{-1 / 2}\right)^{-1}$ exist. Hence, T is nonsingular. However, T is not orthogonal, even if Q is, because each element of each eigenvector is multiplied by the reciprocal of the square root of each eigenvalue, e.g., for the $j^{\text {th }}$ eigenvector of A, i.e., $q_{j}$, the
product $\mathrm{D}^{-1 / 2} \mathrm{q}_{\mathrm{j}}=\mathrm{t}_{\mathrm{j}}$ is:

$$
D^{-1 / 2} q_{j}=\left[\begin{array}{ccc}
\frac{1}{\sqrt{\lambda_{1}}} & & \\
& \frac{1}{\sqrt{\lambda_{2}}} & \\
& & \ddots
\end{array}\right]\left[\begin{array}{c}
q_{1 \mathrm{j}} \\
q_{2 \mathrm{j}} \\
\vdots
\end{array}\right]=\left[\begin{array}{c}
\frac{q_{1 \mathrm{j}}}{\sqrt{\lambda_{1}}} \\
\frac{q_{2 \mathrm{j}}}{\sqrt{\lambda_{2}}} \\
\vdots \\
\end{array}\right]=t_{\mathrm{j}}
$$

Thus,

$$
\mathrm{t}_{\mathrm{j}}^{\prime} \mathrm{t}_{\mathrm{j}}=\sum_{\mathrm{i}=1}^{\mathrm{n}}\left[\frac{\left(\mathrm{q}_{\mathrm{ij}}\right)^{2}}{\lambda_{\mathrm{i}}}\right] \neq 1
$$

and

$$
\mathrm{t}^{\prime} \mathrm{t}_{\mathrm{j}^{\prime}}=\sum_{\mathrm{i}=1}^{\mathrm{n}}\left[\frac{\left(\mathrm{q}_{\mathrm{ij}} \mathrm{q}_{\mathrm{ij}}\right)^{\prime}}{\lambda_{\mathrm{i}}}\right]=0
$$

Thus, $\mathrm{A}=\mathrm{T}^{-1}\left(\mathrm{~T}^{\prime}\right)^{-1}=\mathrm{P}^{\prime} \mathrm{P}$ for $\mathrm{P}=\mathrm{T}^{-1}=\mathrm{Q}^{-1} \mathrm{D}^{1 / 2}$.
(b) If $\mathrm{A}=\mathrm{P}^{\prime} \mathrm{P}$ for P nonsingular, then A is symmetric and

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

which is the sum of squares of Px. Thus,

$$
x^{\prime} A x>0 \text { for all } P x \neq 0
$$

and

$$
x^{\prime} A x=0 \text { for all } P x=0
$$

But $\mathrm{Px}=0$ only when $\mathrm{x}=0$ because P is non-singular, which implies that $\mathrm{P}^{-1}$ exists. Thus,

$$
x^{\prime} A x>0 \quad \text { for all } x \neq 0
$$

and

$$
x^{\prime} A x=0 \quad \text { only for } x=0
$$

$\Rightarrow \quad$ by definition A is p.d.
(9) If $A_{m \times n}$ has full column rank, i.e., the rank $(A)=n$, then $A^{\prime} A$ is positive definite.

Proof: $x^{\prime} A^{\prime} A x$ is the sum of squares of the elements of $A x$. If $A$ is full column rank, then $A x=0$ only when $\mathrm{x}=0$. Thus,

$$
x^{\prime} A^{\prime} A x>0 \text { for all } x \neq 0
$$

$$
\Rightarrow \quad \mathrm{A}^{\prime} \mathrm{A} \text { is } \mathrm{p} . \mathrm{d} .
$$

Corollary: If $\mathrm{A}_{\mathrm{m} \times \mathrm{n}}$ has full row rank, i.e., the $\operatorname{rank}(\mathrm{A})=m$, then $\mathrm{AA}^{\prime}$ is p.d.
(10) The sum of p.s.d. matrices is also p.s.d.

Proof: Let $\mathrm{A}_{\mathrm{i}}, \mathrm{i}=1, \ldots, \mathrm{p}$ be a set of p .s.d. matrices. Then, consider:

$$
x^{\prime}\left(\sum_{i=1}^{p} A_{i}\right) x=x^{\prime} A_{1} x+\ldots+x^{\prime} A_{p} x
$$

Each one of the quadratics $x^{\prime} A_{i} x, i=1, \ldots, p$, is p.s.d. $\Rightarrow$ their sum is positive $\Rightarrow$ the sum of p.d.
matrices is also p.d.

## Properties of positive semi-definite (p.s.d.) matrices

(1) A symmetric matrix A is p.s.d. if and only if all the eigenvalues are either zero or positive with at least one of them equal to zero.
(2) If $A_{n \times n}$ is p.s.d., then,
(a) $|\mathrm{A}|=0$,
(b) $\operatorname{rank}(\mathrm{A})=\mathrm{r}<\mathrm{n}$,
(c) A is singular.
(3) If $A_{n \times n}$ is p.s.d. and $P$ is an $n \times m$ matrix with rank $(P)=m$, then $P^{\prime} A P$ is p.s.d.

## Specializations of property (3):

(3.1) If $A$ is p.s.d. and $P$ is nonsingular, then $P^{\prime} A P$ is p.s.d.
(3.2) If $A$ is p.s.d. then $A^{-}$is p.s.d.
(3.3) If P is an $\mathrm{n} \times \mathrm{m}$ matrix with $\operatorname{rank}(\mathrm{P})=\mathrm{r}<\mathrm{m}$, then $\mathrm{P}^{\prime} \mathrm{P}$ is p.s.d.
(4) If A is p.s.d., then some principal submatrices of A are p.s.d. while others are p.d.
(5) If A is p.s.d., then some principal minors of A are positive while others are zero. In particular,
(a) $\mathrm{a}_{\mathrm{ii}} \geq 0$ for all i with at least one i for which $\mathrm{a}_{\mathrm{ii}}=0$, and
(b) $\mathrm{a}_{\mathrm{ii}} \mathrm{a}_{\mathrm{ij}}-\mathrm{a}_{\mathrm{ij}}{ }^{2} \geq 0$ for all i and j , except for at least one i and j where $\mathrm{a}_{\mathrm{ii}} \mathrm{a}_{\mathrm{ij}}-\mathrm{a}_{\mathrm{ij}}^{2}=0$.
(6) If $A_{n \times n}$ is p.s.d. of rank $r$, there exists a singular matrix $P_{n \times n}$ of rank $r$, such that,
(a) $\mathrm{PAP}^{\prime}=\left[\begin{array}{ll}\mathrm{I}_{\mathrm{r}} & 0 \\ 0 & 0\end{array}\right]$, and
(b) $\mathrm{P}^{\prime} \mathrm{P}=\mathrm{A}^{-}$.

Proof:
(a) $\mathrm{E}^{\prime} \mathrm{AE}=\left[\begin{array}{cc}\mathrm{D}_{\mathrm{r}} & 0 \\ 0 & 0\end{array}\right]$

$$
=D_{n} \quad \text { for } E \text { orthogonal. }
$$

Define:

$$
\mathrm{T}=\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}}^{1 / 2} & 0 \\
0 & 0
\end{array}\right]
$$

Then,

$$
\mathrm{P}=\mathrm{T}^{\prime} \mathrm{E}^{\prime} \Rightarrow \mathrm{P} \text { is singular because } \mathrm{T} \text { is singular. }
$$

Thus,

$$
\begin{aligned}
& \text { PAP }^{\prime}=\mathrm{T}^{\prime} \mathrm{E}^{\prime} A E T \\
& \text { PAP }^{\prime}=\mathrm{T}^{\prime}\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}} & 0 \\
0 & 0
\end{array}\right] \mathrm{T} \\
& \text { PAP }^{\prime}=\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}}^{-1 / 2} & 0 \\
0 & 0
\end{array}\right]\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}} & 0 \\
0 & 0
\end{array}\right]\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}}^{-1 / 2} & 0 \\
0 & 0
\end{array}\right] \\
& \text { PAP }^{\prime}=\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}}^{-1 / 2} \mathrm{D}_{\mathrm{r}}^{1 / 2} \mathrm{D}_{\mathrm{r}}^{1 / 2} \mathrm{D}_{\mathrm{r}}^{-1 / 2} & 0 \\
0 & 0
\end{array}\right] \\
& \text { PAP }^{\prime}=\left[\begin{array}{ll}
\mathrm{I}_{\mathrm{r}} & 0 \\
0 & 0
\end{array}\right]
\end{aligned}
$$

(b) A g-inverse for A must satisfy $\mathrm{AA}^{-} \mathrm{A}=\mathrm{A}$, where $\mathrm{A}=\mathrm{ED}_{\mathrm{n}} \mathrm{E}^{\prime}$, for E orthogonal.

Proof: Consider

$$
\begin{aligned}
& \mathrm{A}^{-}=\left(E D_{\mathrm{n}} \mathrm{E}^{\prime}\right)^{-} \\
& \mathrm{A}^{-}=E \mathrm{D}_{\mathrm{n}}^{-} \mathrm{E}^{\prime}
\end{aligned}
$$

Thus,

$$
\begin{aligned}
\mathrm{AA}^{-} \mathrm{A} & =\left(\mathrm{ED}_{\mathrm{n}} \mathrm{E}^{\prime}\right)\left(\mathrm{ED}_{\mathrm{n}}{ }^{-} \mathrm{E}^{\prime}\right)\left(\mathrm{ED}_{\mathrm{n}} \mathrm{E}^{\prime}\right) \\
& =\mathrm{ED}_{\mathrm{n}} \mathrm{I}_{\mathrm{n}}{ }^{-} \mathrm{ID}_{\mathrm{n}} \mathrm{E}^{\prime} \\
& =\mathrm{ED}_{\mathrm{n}} \mathrm{E}^{\prime} \\
\Rightarrow \quad \mathrm{A}^{-} & =\mathrm{ED}_{\mathrm{n}}{ }^{-} \mathrm{E}^{\prime} \text { is a g-inverse of } \mathrm{A} .
\end{aligned}
$$

But

$$
\begin{aligned}
& \mathrm{D}_{\mathrm{n}}=\mathrm{D}_{\mathrm{n}}^{1 / 2} \mathrm{D}_{\mathrm{n}}^{1 / 2}=\mathrm{TT}^{\prime}=\mathrm{T}^{\prime} \mathrm{T}, \\
\Rightarrow \quad & \mathrm{~A}^{-}=\mathrm{ETT}^{\prime} \mathrm{E}^{\prime}
\end{aligned}
$$

$$
\mathrm{A}^{-}=\mathrm{P}^{\prime} \mathrm{P}
$$

$$
\Rightarrow \quad \mathrm{P}^{\prime} \mathrm{P} \text { is a g-inverse of } \mathrm{A} .
$$

(7) If $A_{n \times n}$ is p.s.d. of rank $r$, there is a full column rank $n \times r$ matrix $L$ such that $A=L L^{\prime}$.

Proof:

$$
\begin{aligned}
& \mathrm{PAP}^{\prime}=\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}} & 0 \\
0 & 0
\end{array}\right] \text { for P orthogonal } \\
& \mathrm{PAP}^{\prime}=\left[\begin{array}{c}
\mathrm{D}_{\mathrm{r}}^{1 / 2} \\
0
\end{array}\right]\left[\begin{array}{ll}
\mathrm{D}_{\mathrm{r}}^{1 / 2} & 0
\end{array}\right]
\end{aligned}
$$

Thus,

$$
\begin{aligned}
& \mathrm{A}=\mathrm{P}^{\prime}\left[\begin{array}{c}
\mathrm{D}_{\mathrm{r}}^{1 / 2} \\
0
\end{array}\right]\left[\begin{array}{ll}
\mathrm{D}_{\mathrm{r}}^{1 / 2} & 0
\end{array}\right] \mathrm{P} \\
& \mathrm{~A}=\mathrm{LL}^{\prime}
\end{aligned}
$$

where

$$
L=P^{\prime}\left[\begin{array}{c}
D_{r}^{1 / 2} \\
0
\end{array}\right] \quad \text { is } n \times r \text { of full column rank, }
$$

and

$$
L^{\prime}=\left[\begin{array}{ll}
D_{r}^{1 / 2} & 0
\end{array}\right] \mathrm{P} \quad \text { is } r \times n \text { of full row rank. }
$$

Also, note that
$\mathrm{L}^{\prime} \mathrm{L}=\left[\begin{array}{ll}\mathrm{D}_{\mathrm{r}}^{1 / 2} & 0\end{array}\right] \mathrm{PP},\left[\begin{array}{c}\mathrm{D}_{\mathrm{r}}^{1 / 2} \\ 0\end{array}\right]$
$L^{\prime} \mathrm{L}=\mathrm{D}_{\mathrm{r}}^{1 / 2} \mathrm{D}_{\mathrm{r}}^{1 / 2}$
$L^{\prime} \mathrm{L}=\mathrm{D}_{\mathrm{r}}$
(8) A symmetric matrix is p.s.d. if it can be written as $\mathrm{P}^{\prime} \mathrm{P}$ for a singular matrix P .

Proof:
(a) Necessary condition: existence of P .

Because A is symmetric,

$$
\begin{aligned}
\mathrm{QAQ}^{\prime} & =\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}} & 0 \\
0 & 0
\end{array}\right] \equiv \mathrm{D}_{\mathrm{n}} \quad \text { for } \mathrm{Q} \text { orthogonal } \\
\Rightarrow \quad \mathrm{A} & =\mathrm{Q}^{\prime} \mathrm{D}_{\mathrm{n}} \mathrm{Q} \\
\mathrm{~A} & =\mathrm{Q}^{\prime} \mathrm{DDQ}
\end{aligned}
$$

where

$$
\begin{aligned}
\mathrm{D} & =\left[\begin{array}{cc}
\mathrm{D}_{\mathrm{r}}^{1 / 2} & 0 \\
0 & 0
\end{array}\right] \\
\Rightarrow \quad \mathrm{A} & =\mathrm{P}^{\prime} \mathrm{P} \text { for } \mathrm{P}=\mathrm{DQ}
\end{aligned}
$$

(b) If $\mathrm{A}=\mathrm{P}^{\prime} \mathrm{P}$ for P singular, then A is symmetric and $\mathrm{x}^{\prime} \mathrm{Ax}=\mathrm{x}^{\prime} \mathrm{P}^{\prime} \mathrm{Px}$, which is the sum of squares of Px . Thus, $\mathrm{x}^{\prime} \mathrm{Ax} \geq 0$ for all $\mathrm{Px} \neq 0$ with at least one $\mathrm{Px} \neq 0$ for which $\mathrm{x}^{\prime} \mathrm{Ax}=0$. But $\mathrm{Px}=0$ at least for one $x \neq 0 \Rightarrow P$ is singular. Hence, $x^{\prime} A x \geq 0$ for all $x \neq 0$ with at least one $x \neq 0$ for which $x^{\prime} A^{\prime} x$ $=0$. So, by definition A is p.s.d.
(9) If $A_{m \times n}$ does not have full column rank, i.e., $\operatorname{rank}(A)=r<m$, then $A A^{\prime}$ is p.s.d.
(10) The sum of p.s.d. matrices is also p.s.d.

Similar theorems to those described above can also be made for n.n.d, n.d., n.s.d. and n.p.d. matrices. In particular, note that if A is n.d., the "nested" principal minors of A alternate in sign, i.e., $a_{i i}<0, a_{i i} a_{j j}-a_{i j}^{2}>0 \ldots$

## Orthogonal matrices

A matrix A is orthogonal if $\mathrm{AA}^{\prime}=\mathrm{I}$, which implies that $\mathrm{A}^{\prime}=\mathrm{A}^{-1}$ and that $\mathrm{A}^{\prime} \mathrm{A}=\mathrm{I}$.

## Properties of orthogonal matrices:

(1) The inner product of any row (column) with itself is 1 , and with any other row (column) is zero.

Proof: This is a consequence of $\mathrm{AA}^{\prime}=\mathrm{I}$.
(2) A product of orthogonal matrices is itself orthogonal.

Proof: Let A and B be two orthogonal matrices. Then,

$$
\begin{aligned}
(\mathrm{AB})(\mathrm{AB})^{\prime} & =\mathrm{ABB}^{\prime} \mathrm{A}^{\prime} \\
& =\mathrm{AIA}^{\prime} \\
& =\mathrm{II} \\
& =\mathrm{I}
\end{aligned}
$$

(3) The determinant of an orthogonal matrix is either 1 or -1 .

Proof: For A orthogonal,

$$
\begin{aligned}
\left|\mathrm{AA}^{\prime}\right| & =|\mathrm{I}| \\
|\mathrm{A}||\mathrm{A}|^{\prime} & =|\mathrm{I}|
\end{aligned}
$$

Thus,

$$
\begin{aligned}
|\mathrm{A}| & =\left|\mathrm{A}^{\prime}\right| \\
\Rightarrow \quad|\mathrm{A}||\mathrm{A}| & =1
\end{aligned}
$$

But $(-1)(-1)=1 \quad$ or $\quad(1)(1)=1$
$\Rightarrow \quad|\mathrm{A}|=1$ or -1
(4) If $\lambda$ is a latent root of an orthogonal matrix A , then so is $\frac{1}{\lambda}$.

Proof:

$$
\begin{aligned}
|\mathrm{A}-\lambda \mathrm{I}| & =\left|\mathrm{AA}^{\prime}-\lambda \mathrm{A}^{\prime}\right|=0 \\
& =\left|\mathrm{I}-\lambda \mathrm{A}^{\prime}\right|=0 \text { for } \mathrm{AA}^{\prime}=\mathrm{I} \\
& =\left|\frac{1}{\lambda} \mathrm{I}-\mathrm{A}^{\prime}\right|=0 \\
& =\left|\frac{1}{\lambda} \mathrm{I}^{\prime}-\mathrm{A}^{\prime}\right|=0 \\
& =\left|\left(\frac{1}{\lambda} \mathrm{I}-\mathrm{A}\right)^{\prime}\right|=0 \\
& =\left|\left(\mathrm{A}-\frac{1}{\lambda} \mathrm{I}\right)^{\prime}\right|=0 \\
& =\left|\mathrm{A}-\frac{1}{\lambda} \mathrm{I}\right|=0
\end{aligned}
$$

## Idempotent Matrices

A matrix $A$ is idempotent if $A^{2}=A$. For instance, the matrix $H=G A$ is idempotent because $(\mathrm{GA})(\mathrm{GA})=\mathrm{G}(\mathrm{AGA})=\mathrm{GA}$.

## Properties of Idempotent Matrices

(1) Idempotent matrices are square.

Proof: A idempotent $\Rightarrow \mathrm{AA}=\mathrm{A}^{2}$ exists only if A is square.
(2) The only nonsingular idempotent matrix is I.

Proof: Consider a nonsingular A, then

$$
\begin{aligned}
\mathrm{A}^{2} & =\mathrm{A} \\
\mathrm{~A}^{-1} \mathrm{~A}^{2} & =\mathrm{A}^{-1} \mathrm{~A} \\
\mathrm{~A}^{-1} \mathrm{AA} & =\mathrm{I} \\
\mathrm{~A} & =\mathrm{I}
\end{aligned}
$$

(3) If A and B are idempotent so is AB , provided that $\mathrm{AB}=\mathbf{B A}$.

Proof:

$$
\begin{aligned}
(\mathrm{AB})^{2} & =\mathrm{ABAB} \\
& =\mathrm{ABBA} \quad \text { if } \mathrm{AB}=\mathrm{BA} \\
& =\mathrm{AB}^{2} \mathrm{~A} \\
& =\mathrm{ABA} \\
& =\mathrm{AAB} \quad \text { if } \mathrm{BA}=\mathrm{AB} \\
& =\mathrm{A}^{2} \mathrm{~B} \\
& =\mathrm{A}^{2} \mathrm{~B}^{2} \\
& =\mathrm{AB}
\end{aligned}
$$

(4) If P is orthogonal and A is idempotent, $\mathrm{P}^{\prime} \mathrm{AP}$ is idempotent.

Proof:

$$
\begin{aligned}
\left(\mathrm{P}^{\prime} \mathrm{AP}\right)\left(\mathrm{P}^{\prime} \mathrm{AP}\right) & =\mathrm{P}^{\prime} \mathrm{AIAP} \\
& =\mathrm{P}^{\prime} \mathrm{A}^{2} \mathrm{P} \\
& =\mathrm{P}^{\prime} \mathrm{AP}
\end{aligned}
$$

(5) The latent roots of an idempotent matrix are either 0 or 1 .

Proof: Let A be an idempotent matrix with an eigenvalue $\lambda$ and its eigenvector $u$.
Thus,

$$
\begin{aligned}
\mathrm{Au} & =\lambda u \\
\mathrm{~A}^{2} \mathrm{u} & =\lambda^{2} \mathrm{u}
\end{aligned}
$$

But

$$
\begin{aligned}
\mathrm{A}^{2} u & =\mathrm{Au} \\
\Rightarrow \quad \lambda^{2} u & =\lambda u \\
\Rightarrow \quad\left(\lambda^{2}-\lambda\right) u & =0
\end{aligned}
$$

Also, because $u \neq 0$,

$$
\begin{aligned}
& \left(\lambda^{2}-\lambda\right) \quad=0 \\
& \lambda(\lambda-1)=0 \\
\Rightarrow & \lambda_{1}=0 \quad \text { and } \quad \lambda_{2}=1
\end{aligned}
$$

(6) The number of eigenvalues of an idempotent matrix is the same as its rank.

Proof: Let matrix A be idempotent with rank $(\mathrm{A})=\mathrm{r}$. Let D be the equivalent diagonal form of A
whose diagonal elements are the eigenvalues of $A$. Thus, $\operatorname{rank}(D)=\operatorname{rank}(A)=r \Rightarrow$ by property
(5) above, the only nonzero diagonal elements of D are 1 's, and there must be r of them.
(7) The trace of an idempotent matrix is equal to its rank.

Proof: Trace (A) = Trace (D) = r by property (6).
(8) A general form for an idempotent matrix is $\mathrm{A}=\mathrm{X}(\mathrm{YX})^{-1} \mathrm{Y}$ provided that $(\mathrm{YX})^{-1}$ exists.

Proof:

$$
\begin{aligned}
\mathrm{A}^{2} & =\left(\mathrm{X}(\mathrm{YX})^{-1} \mathrm{Y}\right)\left(\mathrm{X}(\mathrm{YX})^{-1} \mathrm{Y}\right) \\
& =\mathrm{X}(\mathrm{YX})^{-1} \mathrm{IY} \\
& =\mathrm{X}(\mathrm{YX})^{-1} \mathrm{Y}
\end{aligned}
$$

(9) A general form for an idempotent symmetric matrix is $A=X\left(X^{\prime} X\right)^{-1} X^{\prime}$, provided that $\left(\mathrm{X}^{\prime} \mathrm{X}\right)^{-1}$ exists.

Proof:

$$
\begin{aligned}
\mathrm{A}^{2} & =\mathrm{X}\left(\mathrm{X}^{\prime} \mathrm{X}\right)^{-1} \mathrm{X}^{\prime} \mathrm{X}\left(\mathrm{X}^{\prime} \mathrm{X}\right)^{-1} \mathrm{X} \\
& =\mathrm{X}\left(\mathrm{X}^{\prime} \mathrm{X}\right)^{-1} \mathrm{IX} \\
& =\mathrm{X}\left(\mathrm{X}^{\prime} \mathrm{X}\right)^{-1} \mathrm{X}
\end{aligned}
$$

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