

Basic concepts in matrix algebra

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2022-03-10

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This appendix supplies some general matrix algebra that is useful in these notes. Proof of these, as well as additional explanation, can be found in e.g. Searle [1982], Searle et al. [1992] and/or Knight [2008]. And Schaeffer's EuropeNotes p. 194.

The matrix V is the square variance-covariance matrix $V = ZGZ' + D$.

Denoting the derivative of the matrix:

$$\mathbf{V}_i = \frac{\partial \mathbf{V}}{\partial \sigma_i^2} = \begin{cases} \mathbf{D}_n & \text{when } \sigma_i^2 = \sigma_e^2 \\ \mathbf{ZGZ}' & \text{when } \sigma_i^2 = \sigma_g^2 \end{cases} \quad (1)$$

The derivative of the first quadratic form with respect to X :

$$\frac{\partial (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \mathbf{V} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})}{\partial \mathbf{X}} = -2\mathbf{X}' \mathbf{V} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta}) \quad (2)$$

The derivative of the *inverse* matrix

$$\frac{\partial \mathbf{V}^{-1}}{\partial \sigma_i^2} = -\mathbf{V}^{-1} \mathbf{V}_i \mathbf{V}^{-1} \quad (3)$$

The derivative of the *trace* of the matrix

$$\frac{\partial \text{Tr}(\mathbf{V})}{\partial \sigma_i^2} = \text{Tr}(\mathbf{V}_i) \quad (4)$$

The derivative of the logarithm of the matrix' determinant:

$$\frac{\partial \ln |\mathbf{V}|}{\partial \sigma_i^2} = \text{Tr}(\mathbf{V}^{-1} \mathbf{V}_i) \quad (5)$$

The derivative of the first and second quadratic forms, when taken with respect to σ_i^2 :

$$\frac{\partial (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \mathbf{V} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})}{\partial \sigma_i^2} = (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \mathbf{V}_i (\mathbf{y} - \mathbf{X}\boldsymbol{\beta}) \quad (6a)$$

$$\frac{\partial (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \mathbf{V}^{-1} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})}{\partial \sigma_i^2} = -(\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \mathbf{V}^{-1} \mathbf{V}_i \mathbf{V}^{-1} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta}) \quad (6b)$$

and the derivative of the logarithm of the determinant of the second quadratic form:

$$\frac{\partial \ln |\mathbf{X}'\mathbf{V}^{-1}\mathbf{X}|}{\partial \sigma_i^2} = \text{Tr}((\mathbf{X}'\mathbf{V}^{-1}\mathbf{X})^{-1}\mathbf{X}'\mathbf{V}^{-1}\mathbf{V}_i\mathbf{V}^{-1}\mathbf{X}) \quad (6c)$$

but also [Knight, 2008, p. 21]:

$$\begin{aligned} & \frac{\partial \ln |\mathbf{V}|}{\partial \sigma_i^2} + \frac{\partial \ln |\mathbf{X}'\mathbf{V}^{-1}\mathbf{X}|}{\partial \sigma_i^2} \\ &= \text{Tr}(\mathbf{V}^{-1}\mathbf{V}_i) - \text{Tr}((\mathbf{X}'\mathbf{V}^{-1}\mathbf{X})^{-1}\mathbf{X}'\mathbf{V}^{-1}\mathbf{V}_i\mathbf{V}^{-1}\mathbf{X}) \\ &= \text{Tr}(\mathbf{V}^{-1}\mathbf{V}_i - (\mathbf{X}'\mathbf{V}^{-1}\mathbf{X})^{-1}\mathbf{X}'\mathbf{V}^{-1}\mathbf{V}_i\mathbf{V}^{-1}\mathbf{X}) \\ &= \text{Tr}(\mathbf{P}\mathbf{V}_i). \end{aligned}$$

A *projection matrix*, \mathbf{P} can be defined as

$$\mathbf{P} = \mathbf{V}^{-1} - \mathbf{V}^{-1}\mathbf{X}(\mathbf{X}'\mathbf{V}^{-1}\mathbf{X})^{-1}\mathbf{X}'\mathbf{V}^{-1} \quad (7)$$

% and has following nice properties:

$$\mathbf{P}\mathbf{X} = 0 \quad (8)$$

$$\mathbf{P}\mathbf{y} = \mathbf{V}^{-1}(\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}}), \text{ where} \quad (9)$$

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}'\mathbf{V}^{-1}\mathbf{X})^{-1}\mathbf{X}'\mathbf{V}^{-1}\mathbf{y} \quad (10)$$

$$\hat{\mathbf{P}}\mathbf{y} = \hat{\mathbf{e}} \quad (11)$$

The derivative of the projection matrix:

$$\frac{\partial \mathbf{P}}{\partial \sigma_i^2} = -\mathbf{P}\mathbf{V}_i\mathbf{P} \quad (12)$$

For traces, we have:

$$\text{Tr}(\mathbf{P}) = \mathbf{y}'\mathbf{P}\mathbf{P}\mathbf{y} \quad (13a)$$

or

$$\text{Tr}(\mathbf{P}\mathbf{V}_i) = \mathbf{y}'\mathbf{P}\mathbf{V}_i\mathbf{P}\mathbf{y}. \quad (13b)$$

%

Some forms

In (??) on page ??, the form $\mathbf{Z}\mathbf{G}\mathbf{Z}'$ is used. The following is the result if we consider a very simple example with two animals, where the first has two observations and the second only one.

$$\begin{aligned} \mathbf{y} &= \mu + \mathbf{Z}\mathbf{g} + \mathbf{e} \\ \mathbf{Z} &= \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} & \mathbf{G} &= \begin{bmatrix} g_{11} & g_{12} \\ g_{21} & g_{22} \end{bmatrix} \\ \mathbf{Z}\mathbf{G}\mathbf{Z}' &= \begin{bmatrix} g_{11} & g_{11} & g_{12} \\ g_{11} & g_{11} & g_{12} \\ g_{21} & g_{21} & g_{22} \end{bmatrix} & \mathbf{Z}\mathbf{Z}' &= \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{aligned} \quad (14)$$

1 Derivation of mean and variance of quadratic forms

If we assume that $\mathbf{y} \sim N(\mu, \mathbf{V})$ then [Searle et al., 1992, p. 466]

$$\begin{aligned} E(\mathbf{y}'\mathbf{Q}\mathbf{y}) &= \text{Tr}(\mathbf{Q}\mathbf{V}) + \mu'\mathbf{Q}\mu \\ \text{Var}(\mathbf{y}'\mathbf{Q}\mathbf{y}) &= 2 \text{Tr}(\mathbf{Q}\mathbf{V}\mathbf{Q}\mathbf{V}) + 4\mu'\mathbf{Q}\mathbf{V}\mathbf{Q}\mu \end{aligned}$$

The quadratic form $\mathbf{y}'\mathbf{Q}\mathbf{y}$ has a non-central χ^2 distribution with $\text{rank}(\mathbf{Q})$ degrees of freedom and non-central parameter $\frac{1}{2}\mu'\mathbf{Q}\mu$, if and only if $\mathbf{Q}\mathbf{V}$ is idempotent (Theorem S2, p. 467, Searle et al., 1992). When $\mathbf{Q}\mathbf{V}$ is idempotent, then $\mathbf{Q}\mathbf{V}\mathbf{Q}\mathbf{V} = \mathbf{Q}\mathbf{V} \Rightarrow \text{Tr}(\mathbf{Q}\mathbf{V}\mathbf{Q}\mathbf{V}) = \text{Tr}(\mathbf{Q}\mathbf{V})$, and $\text{Tr}(\mathbf{Q}\mathbf{V}) = \text{rank}(\mathbf{Q}\mathbf{V})$.

Because $\mu = 0$, $\mathbf{V} = \mathbf{I}$, and $\mathbf{Q} = \mathbf{V}_0^{-\frac{1}{2}}\mathbf{M}\mathbf{V}_0^{\frac{1}{2}}$ (where $\mathbf{M} = \mathbf{V}_0^{-1}\mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1}$), the expectation and variance of the quadratic form are

$$\begin{aligned} E(\mathbf{y}'\mathbf{Q}\mathbf{y}) &= \text{Tr}(\mathbf{Q}) \\ &= \text{Tr} \left(\frac{1}{2} \mathbf{V}_0^{\frac{1}{2}} \mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{V}_0^{\frac{1}{2}} \right) \\ &= \frac{1}{2} \text{Tr} \left(\mathbf{V}_0^{\frac{1}{2}} \mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{V}_0^{\frac{1}{2}} \right) \\ &= \frac{1}{2} \text{Tr} (\mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}') \end{aligned}$$

and

$$\begin{aligned} \text{Var}(\mathbf{y}'\mathbf{Q}\mathbf{y}) &= 2 \text{Tr}(\mathbf{Q}\mathbf{Q}) \\ &= 2 \text{Tr} \left(\left[\frac{1}{2} \mathbf{V}_0^{\frac{1}{2}} \mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{V}_0^{\frac{1}{2}} \right] \left[\frac{1}{2} \mathbf{V}_0^{\frac{1}{2}} \mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{V}_0^{\frac{1}{2}} \right] \right) \\ &= \frac{1}{2} \text{Tr} \left(\left[\mathbf{V}_0^{\frac{1}{2}} \mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{V}_0^{\frac{1}{2}} \right] \left[\mathbf{V}_0^{\frac{1}{2}} \mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{V}_0^{\frac{1}{2}} \right] \right) \\ &= \frac{1}{2} \text{Tr} \left(\mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{V}_0^{\frac{1}{2}} \mathbf{V}_0^{\frac{1}{2}} \right) \\ &= \frac{1}{2} \text{Tr} \left(\mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}'\mathbf{V}_0^{-1} \mathbf{Z}\mathbf{G}_i\mathbf{Z}' \right). \end{aligned}$$

We can further elaborate that $\mathbf{y}'\mathbf{Q}\mathbf{y}$ is χ^2 distributed with $\text{Tr}(\mathbf{Q})$ degrees of freedom under the conditions above.

References

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