

Some useful Gaussian and matrix equations

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Appendix: Some useful Gaussian identities

If \mathbf{x} is multivariate Gaussian with mean $\boldsymbol{\mu}$ and covariance matrix $\boldsymbol{\Sigma}$

$$p(\mathbf{x}; \boldsymbol{\mu}, \boldsymbol{\Sigma}) = (2\pi|\boldsymbol{\Sigma}|)^{-D/2} \exp\left(-(\mathbf{x} - \boldsymbol{\mu})^\top \boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu})/2\right),$$

then

$$\mathbb{E}[\mathbf{x}] = \boldsymbol{\mu},$$

$$\mathbb{V}[\mathbf{x}] = \mathbb{E}[(\mathbf{x} - \mathbb{E}[\mathbf{x}])^2] = \boldsymbol{\Sigma}.$$

For any matrix A , if $\mathbf{z} = A\mathbf{x}$ then \mathbf{z} is Gaussian and

$$\mathbb{E}[\mathbf{z}] = A\boldsymbol{\mu},$$

$$\mathbb{V}[\mathbf{z}] = A\boldsymbol{\Sigma}A^\top.$$

Matrix and Gaussian identities cheat sheet

Matrix identities

- Matrix inversion lemma (Woodbury, Sherman & Morrison formula)

$$(Z + UWV^T)^{-1} = Z^{-1} - Z^{-1}U(W^{-1} + V^T Z^{-1}U)^{-1}V^T Z^{-1}$$

- A similar equation exists for determinants

$$|Z + UWV^T| = |Z| |W| |W^{-1} + V^T Z^{-1}U|$$

The product of two Gaussian density functions

$$\mathcal{N}(\mathbf{x}|\mathbf{a}, \mathbf{A}) \mathcal{N}(\mathbf{P}^T \mathbf{x}|\mathbf{b}, \mathbf{B}) = z_c \mathcal{N}(\mathbf{x}|\mathbf{c}, \mathbf{C})$$

- is proportional to a Gaussian density function with covariance and mean

$$\mathbf{C} = (\mathbf{A}^{-1} + \mathbf{P} \mathbf{B}^{-1} \mathbf{P}^T)^{-1} \quad \mathbf{c} = \mathbf{C} (\mathbf{A}^{-1} \mathbf{a} + \mathbf{P} \mathbf{B}^{-1} \mathbf{b})$$

- and has a normalizing constant z_c that is Gaussian both in \mathbf{a} and in \mathbf{b}

$$z_c = (2\pi)^{-\frac{m}{2}} |\mathbf{B} + \mathbf{P}^T \mathbf{A} \mathbf{P}|^{-\frac{1}{2}} \exp\left(-\frac{1}{2}(\mathbf{b} - \mathbf{P}^T \mathbf{a})^T (\mathbf{B} + \mathbf{P}^T \mathbf{A} \mathbf{P})^{-1} (\mathbf{b} - \mathbf{P}^T \mathbf{a})\right)$$