

# Some useful Gaussian and matrix equations

Carl Edward Rasmussen

June 23rd, 2016

# Appendix: Some useful Gaussian identities

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

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

then

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

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

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

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

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

# Matrix and Gaussian identities cheat sheet

## Matrix identities

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

$$(Z + U W V^\top)^{-1} = Z^{-1} - Z^{-1} U (W^{-1} + V^\top Z^{-1} U)^{-1} V^\top Z^{-1}$$

- A similar equation exists for determinants

$$|Z + U W V^\top| = |Z| |W| |W^{-1} + V^\top Z^{-1} U|$$

The product of two Gaussian density functions

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

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

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

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

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