Tutorial Week 8

COMP9418 – Advanced Topics in Statistical Machine Learning, 18s2, UNSW Sydney

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Last Update: Tuesday 4th September, 2018 at 15:47

- 1. (Bishop, PRML, 2006) Let x be a D-dimensional random variable having a Gaussian distribution given by $N(\mathbf{x}|\boldsymbol{\mu},\boldsymbol{\Sigma})$, and consider the M-dimensional random variable given by $\mathbf{y} = \mathbf{A}\mathbf{x} + \mathbf{b}$ where **A** is an $M \times D$ matrix. Show that \mathbf{y} also has a Gaussian distribution, and find expressions for its mean and covariance. Discuss the form of this Gaussian distribution for M < D, for M = D, and for M > D.
- 2. (Original version from Bishop, PRML, 2006) Suppose we replace the zero-mean, unitcovariance latent space prior $p(\mathbf{z}) = \mathcal{N}(\mathbf{z}|\mathbf{0}, \mathbf{I})$ in the PPCA model by a general Gaussian distribution of the form $p(\mathbf{z}) = \mathcal{N}(\mathbf{z}|\mathbf{m}_z, \Sigma_z)$. By redefining the parameters of the model, show that this leads to an identical model for the marginal distribution $p(\mathbf{x})$ over the observed variables for any valid choice of \mathbf{m}_z and Σ_z .
- 3. Show that, as in the Factor Analysis model, given the Gaussian prior $p(\mathbf{z}) = \mathcal{N}(\mathbf{z}|\mathbf{0},\mathbf{I})$ and the conditional likelihood $p(\mathbf{x}|\mathbf{z}) = \mathcal{N}(\mathbf{x}|\mathbf{W}\mathbf{z} + \boldsymbol{\mu}, \boldsymbol{\Psi})$, the marginal likelihood is given by $p(\mathbf{x}) = \mathcal{N}(\mathbf{x}|\boldsymbol{\mu}, \mathbf{W}\mathbf{W}^T + \boldsymbol{\Psi}).$
- 4. (Bishop, PRML, 2006) Show that the factor analysis model is invariant under rotations of the latent space coordinates.
- 5. (Original version from Barber, BRML, 2017) Show that under the ICA model given by:

$$p(\mathbf{z}) = \prod_{k=1}^{K} p_k(z_k),$$

$$p(\mathbf{x}|\mathbf{z}, \mathbf{W}, \sigma^2) = \mathcal{N}(\mathbf{x}|\mathbf{W}\mathbf{z}, \sigma^2\mathbf{I}),$$
(1)

$$p(\mathbf{x}|\mathbf{z}, \mathbf{W}, \sigma^2) = \mathcal{N}(\mathbf{x}|\mathbf{W}\mathbf{z}, \sigma^2\mathbf{I}),$$
 (2)

where each individual source distribution $p_k(z_k)$ is non-Gaussian, the posterior $p(\mathbf{z}|\mathbf{x},\mathbf{W},\sigma^2)$ is non-factorised, non-Gaussian and generally intractable (its normalisation constant cannot be computed efficiently).