PRML Errata

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Preface

This report communicates some possible errata for PRML (Bishop, 2006) that are not listed in the official errata document (Svensén and Bishop, 2011)¹ at the time of this writing. When specifying the location of an error, I follow the notational conventions adopted by Svensén and Bishop (2011). I have also included in this report some suggestions for improving the readability.

Corrections

Page 51

Equation (1.98): Following the notation (1.93), we should write the left hand side of (1.98) as H[X] instead of H[p].

Page 80

Equation (2.52): We usually take eigenvectors \mathbf{u}_i to be the columns of \mathbf{U} as in (C.37). If we follow this convention, Equation (2.52) and the following text should read

$$\mathbf{y} = \mathbf{U}^{\mathrm{T}}(\mathbf{x} - \boldsymbol{\mu}) \tag{1}$$

where **U** is a matrix whose columns are given by \mathbf{u}_i so that $\mathbf{U} = (\mathbf{u}_1, \dots, \mathbf{u}_D)$. From (2.46) it follows that **U** is an *orthogonal* matrix, i.e., it satisfies $\mathbf{U}^T\mathbf{U} = \mathbf{I}$ and hence also $\mathbf{U}\mathbf{U}^T = \mathbf{I}$ where **I** is the identity matrix.

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Equations (2.53) and (2.54): If we write the change of variable from \mathbf{x} to \mathbf{y} as (1) instead of (2.52), the Jacobian matrix $\mathbf{J} = (J_{ij})$ we require is simply given by \mathbf{U} . Equation (2.53) should read

$$J_{ij} = \frac{\partial x_i}{\partial y_j} = U_{ij} \tag{2}$$

where U_{ij} is the (i, j)-th element of the matrix **U**. The square of the determinant of the Jacobian matrix (2.54) can be evaluated as follows.

$$|\mathbf{J}|^{2} = |\mathbf{U}|^{2} = |\mathbf{U}^{\mathrm{T}}||\mathbf{U}| = |\mathbf{U}^{\mathrm{T}}\mathbf{U}| = |\mathbf{I}| = 1.$$
(3)

¹ The last line but one of the bibliographic information page of the copy of PRML I have reads "9 8 7 (corrected at 6th printing 2007)." So I refer to Version 2 of the errata.

Line -1: Since the determinant of the Jacobian, which is an orthonormal matrix here, can be negative, we should write $|\mathbf{J}| = \pm 1$ instead of $|\mathbf{J}| = 1$.

Page 82

Equation (2.56): We should take the absolute value of the determinant for the same reason given above; the factor $|\mathbf{J}|$ should read $|\det(\mathbf{J})|$. Note that we cannot write $||\mathbf{J}||$ to mean $|\det(\mathbf{J})|$ because it is confusingly similar to the matrix norm $||\mathbf{J}||$, which usually refers to the largest singular value of \mathbf{J} (Golub and Van Loan, 2013). This notational inconsistency has been caused by the abuse of the notation $|\cdot|$ for both the absolute value and the matrix determinant. If we always use $\det(\cdot)$ for the determinant, confusion will not arise and the notation be consistent. An alternative solution to this problem would be to explicitly define

$$|\mathbf{A}| \equiv |\det(\mathbf{A})| \tag{4}$$

for any square matrix \mathbf{A} . This notation is mostly consistent because we have $|\mathbf{A}| = \det(\mathbf{A})$ for a positive semidefinite matrix \mathbf{A} and most other matrices for which we take determinants are positive (semi)definite in PRML.

Page 104

The text after Equation (2.160): The Gaussian $\mathcal{N}(\mathbf{x}|\boldsymbol{\mu},\boldsymbol{\Lambda})$ should read $\mathcal{N}(\mathbf{x}|\boldsymbol{\mu},\boldsymbol{\Lambda}^{-1})$.

Page 141

Equation (3.13): The use of the gradient operator is not consistent here. As in Equation (2.224), the gradient (of a scalar function) is a column vector so that Equation (3.13) should read²

$$\nabla \ln p\left(\mathbf{t}|\mathbf{w},\beta\right) = \beta \sum_{n=1}^{N} \left\{ t_n - \mathbf{w}^{\mathrm{T}} \boldsymbol{\phi}\left(\mathbf{x}_n\right) \right\} \boldsymbol{\phi}\left(\mathbf{x}_n\right).$$
 (5)

Moreover, I would like to also suggest that we should give a definition for the gradient before we use it or in an appendix. Although Appendix C defines the vector derivative $\frac{\partial}{\partial \mathbf{x}}$, which is used interchangeably with the gradient $\nabla_{\mathbf{x}}$ in PRML, it has no mention of the gradient.

Page 142

Equation (3.14): The left hand side should be a zero vector $\mathbf{0}$ instead of a scalar zero 0. Thus, Equation (3.14) should read

$$\mathbf{0} = \sum_{n=1}^{N} t_n \boldsymbol{\phi}(\mathbf{x}_n) - \left(\sum_{n=1}^{N} \boldsymbol{\phi}(\mathbf{x}_n) \boldsymbol{\phi}(\mathbf{x}_n)^{\mathrm{T}}\right) \mathbf{w}$$
 (6)

where we have used the gradient of the form (5) instead of (3.13).

Page 146

Equation (3.31): The left hand side should be y(x, W) instead of y(x, w).

² I have not got a right typeface for the data vector $(t_1, \ldots, t_N)^{\mathrm{T}}$.

The second paragraph, Line 1: 'Gamma' should read 'gamma' (without capitalization).

Pages 168–169, and 177

Equations (3.88), (3.93), and (3.117) as well as the text before (3.93): The derivative operators should be partial differentials. For example, Equation (3.117) should read

$$\frac{\partial}{\partial \alpha} \ln |\mathbf{A}| = \operatorname{Tr} \left(\mathbf{A}^{-1} \frac{\partial}{\partial \alpha} \mathbf{A} \right). \tag{7}$$

Page 207

Equation (4.92): On the right hand side, the gradient and the Hessian, which are in general functions of the parameter \mathbf{w} , must be evaluated at the previous estimate \mathbf{w}^{old} for the parameter. Thus, Equation (4.92) should read

$$\mathbf{w}^{\text{new}} = \mathbf{w}^{\text{old}} - \left[\mathbf{H} \left(\mathbf{w}^{\text{old}} \right) \right]^{-1} \nabla E \left(\mathbf{w}^{\text{old}} \right)$$
(8)

where $\mathbf{H}(\mathbf{w}) \equiv \nabla \nabla E(\mathbf{w})$ is the Hessian matrix whose elements comprise the second derivatives of $E(\mathbf{w})$ with respect to the components of \mathbf{w} .

Page 210

Equation (4.110): The left hand side of (4.110), which is obtained by taking the gradient of $\nabla_{\mathbf{w}_j} E$ given in (4.109) with respect to \mathbf{w}_k , refers to the (k, j)-th block of the Hessian, not the (j, k)-th. Thus, Equation (4.110) should read

$$\nabla_{\mathbf{w}_k} \nabla_{\mathbf{w}_j} E\left(\mathbf{w}_1, \dots, \mathbf{w}_K\right) = \sum_{n=1}^N y_{nj} \left(I_{kj} - y_{nk}\right) \boldsymbol{\phi}_n \boldsymbol{\phi}_n^{\mathrm{T}}.$$
 (9)

To be clear, we have used the following notation. If we group all the parameters $\mathbf{w}_1, \ldots, \mathbf{w}_K$ into a column vector

$$\mathbf{w} = \begin{pmatrix} \mathbf{w}_1 \\ \vdots \\ \mathbf{w}_K \end{pmatrix} \tag{10}$$

the gradient and the Hessian of the error function $E(\mathbf{w})$ with respect to \mathbf{w} are given by

$$\nabla_{\mathbf{w}} E = \begin{pmatrix} \nabla_{\mathbf{w}_1} E \\ \vdots \\ \nabla_{\mathbf{w}_K} E \end{pmatrix}, \qquad \nabla_{\mathbf{w}} \nabla_{\mathbf{w}} E = \begin{pmatrix} \nabla_{\mathbf{w}_1} \nabla_{\mathbf{w}_1} E & \cdots & \nabla_{\mathbf{w}_1} \nabla_{\mathbf{w}_K} E \\ \vdots & \ddots & \vdots \\ \nabla_{\mathbf{w}_K} \nabla_{\mathbf{w}_1} E & \cdots & \nabla_{\mathbf{w}_K} \nabla_{\mathbf{w}_K} E \end{pmatrix}$$
(11)

respectively.

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Figure 5.6: The eigenvectors \mathbf{u}_1 and \mathbf{u}_2 in Figure 5.6 are unit vectors; their orientations should be shown as in Figure 2.7. Or, the scaled vectors should be labeled as $\lambda_1^{-1/2}\mathbf{u}_1$ and $\lambda_2^{-1/2}\mathbf{u}_2$.

The second paragraph: The approximation of the form (5.84) is usually referred to as the "Gauss-Newton" approximation, but not "Levenberg-Marquardt." The Levenberg-Marquardt method is a method that improves the numerical stability of (Gauss-)Newton iterations by correcting the Hessian matrix so as to be more diagonal dominant (Press et al., 1992).

Page 275

The text after Equation (5.154): The identity matrix I should multiply $\sigma_k^2(\mathbf{x}_n)$.

Page 277

Equation (5.160): The factor L should multiply $\sigma_k^2(\mathbf{x})$ because we have

$$s^{2}(\mathbf{x}) = \mathbb{E}\left[\operatorname{Tr}\left\{\left(\mathbf{t} - \mathbb{E}\left[\mathbf{t}|\mathbf{x}\right]\right)\left(\mathbf{t} - \mathbb{E}\left[\mathbf{t}|\mathbf{x}\right]\right)^{\mathrm{T}}\right\}\middle|\mathbf{x}\right]$$
(12)

$$= \sum_{k=1}^{K} \pi_k(\mathbf{x}) \operatorname{Tr} \left\{ \sigma_k^2(\mathbf{x}) \mathbf{I} + (\boldsymbol{\mu}_k(\mathbf{x}) - \mathbb{E} \left[\mathbf{t} | \mathbf{x} \right]) (\boldsymbol{\mu}_k(\mathbf{x}) - \mathbb{E} \left[\mathbf{t} | \mathbf{x} \right])^{\mathrm{T}} \right\}$$
(13)

$$= \sum_{k=1}^{K} \pi_k(\mathbf{x}) \left\{ L \sigma_k^2(\mathbf{x}) + \|\boldsymbol{\mu}_k(\mathbf{x}) - \mathbb{E}\left[\mathbf{t}|\mathbf{x}\right]\|^2 \right\}$$
(14)

where L is the dimensionality of \mathbf{t} .

Page 295

Line 1: The vector \mathbf{x} should be a column vector so that $\mathbf{x} = (x_1, x_2)^{\mathrm{T}}$.

Page 318

The text before Equation (6.93) as well as Equations (6.93) and (6.94): The text and the equations should read: We can evaluate the derivative of a_n^* with respect to θ_j by differentiating the relation (6.84) with respect to θ_j to give

$$\frac{\partial \mathbf{a}_{N}^{\star}}{\partial \theta_{i}} = \frac{\partial \mathbf{C}_{N}}{\partial \theta_{i}} \left(\mathbf{t}_{N} - \boldsymbol{\sigma}_{N} \right) - \mathbf{C}_{N} \mathbf{W}_{N} \frac{\partial \mathbf{a}_{N}^{\star}}{\partial \theta_{i}}$$
(15)

where the derivatives are Jacobians defined by (C.16) for a vector and analogously for a matrix³. Rearranging (15) then gives

$$\frac{\partial \mathbf{a}_{N}^{\star}}{\partial \theta_{j}} = \left(\mathbf{I} + \mathbf{C}_{N} \mathbf{W}_{N}\right)^{-1} \frac{\partial \mathbf{C}_{N}}{\partial \theta_{j}} \left(\mathbf{t}_{N} - \boldsymbol{\sigma}_{N}\right). \tag{16}$$

Page 355

Equation (7.117): The typeface of the vector \mathbf{y} in (7.117) should be that in (7.110).

Page 414

Figure 8.53, Line 6: The term "max-product" should be "max-sum."

³ The Jacobian $\frac{\partial \mathbf{A}}{\partial \theta}$ of a matrix $\mathbf{A} = (A_{ij})$ with respect to a scalar θ is a matrix with the same dimensionality as \mathbf{A} whose (i,j)-th element is given by $\frac{\partial A_{ij}}{\partial \theta}$.

Equation (9.3): The right hand side should be a zero vector **0** instead of a scalar zero 0.

Page 432

The text after Equation (9.13): It should be noted for clarity that, as the prior $p(\mathbf{z})$ over \mathbf{z} is a multinomial distribution (9.10), the posterior $p(\mathbf{z}|\mathbf{x})$ over \mathbf{z} given \mathbf{x} is again a multinomial of the form

$$p\left(\mathbf{z}|\mathbf{x}\right) = \prod_{k=1}^{K} \gamma_k^{z_k} \tag{17}$$

where we have written $\gamma_k \equiv \gamma(z_k)$, which can be directly confirmed by inspecting the functional form of the joint distribution

$$p(\mathbf{z}) p(\mathbf{x}|\mathbf{z}) = \prod_{k=1}^{K} \left\{ \pi_k \mathcal{N}(\mathbf{x}|\boldsymbol{\mu}_k, \boldsymbol{\Sigma}_k) \right\}^{z_k}.$$
 (18)

This observation helps the reader to understand that evaluating the responsibilities $\gamma(z_k)$ indeed corresponds to the E step of the general EM algorithm.

Page 434

Equation (9.15): Although the official errata (Svensén and Bishop, 2011) states that σ_j on the right hand side should be raised to a power of D, the whole right hand side should be raised to D so that Equation (9.15) should read

$$\mathcal{N}\left(\mathbf{x}_{n}\middle|\mathbf{x}_{n},\sigma_{j}^{2}\mathbf{I}\right) = \frac{1}{\left(2\pi\sigma_{j}^{2}\right)^{D/2}}.$$
(19)

Page 435

Equation (9.16): The right hand side should be a zero vector $\mathbf{0}$.

Page 465

Equations (10.6) and (10.7): In PRML, Equation (10.6) will be later recognized as "a negative Kullback-Leibler divergence between $q_j(\mathbf{Z}_j)$ and $\tilde{p}(\mathbf{X}, \mathbf{Z}_j)$ " (Page 465, Line -2). However, there is no point in taking a Kullback-Leibler divergence between two probability distributions over different sets of random variables; such a quantity is undefined. Moreover, the discussion here seems to be somewhat redundant. We actually do not have to introduce the probability $\tilde{p}(\mathbf{X}, \mathbf{Z}_j)$ other than $q_j^*(\mathbf{Z}_j)$. Specifically, we can rewrite Equations (10.6) and (10.7) into

$$\mathcal{L}(q) = \dots \tag{20}$$

$$= \int q_j \ln q_j^* d\mathbf{Z}_j - \int q_j \ln q_j d\mathbf{Z}_j + \text{const}$$
 (21)

$$= -\operatorname{KL}\left(q_{j} \middle\| q_{j}^{\star}\right) + \operatorname{const} \tag{22}$$

where we have defined a new distribution $q_i^{\star}(\mathbf{Z}_i)$ by the relation

$$\ln q_j^{\star}(\mathbf{Z}_j) = \mathbb{E}_{i \neq j} \left[\ln p\left(\mathbf{X}, \mathbf{Z} \right) \right] + \text{const.}$$
(23)

It directly follows from (22) that, since the lower bound $\mathcal{L}(q)$ is the negative Kullback-Leibler divergence between $q_j(\mathbf{Z}_j)$ and $q_j^{\star}(\mathbf{Z}_j)$ up to some additive constant, the maximum of $\mathcal{L}(q)$ occurs when $q_j(\mathbf{Z}_j) = q_j^{\star}(\mathbf{Z}_j)$.

The text before Equation (10.8): The latent variable \mathbf{z}_i should read \mathbf{Z}_i .

Page 465

Line -1: If we adopt the representation (22), the probability $\tilde{p}(\mathbf{X}, \mathbf{Z}_j)$ should read $q_j^{\star}(\mathbf{Z}_j)$.

Page 466

Line 1: Again, $\tilde{p}(\mathbf{X}, \mathbf{Z}_j)$ should read $q_j^*(\mathbf{Z}_j)$. The sentence "Thus we obtain..." should read "Thus we see that we have already obtained a general expression for the optimal solution in (23)."

Page 468

The text after Equation (10.16): The constant term in (10.16) is the negative entropy of $p(\mathbf{Z})$.

Page 478

Equation (10.63): The additive constant +1 on the right hand side should be omitted so that Equation (10.63) should read

$$\nu_k = \nu_0 + N_k. \tag{24}$$

A quick check for the correctness of the re-estimation equations would be to consider a limit of $N \to 0$, in which the effective number of observations N_k also goes to zero and the re-estimation equations should reduce to identities. Equation (10.63) does not reduce to $\nu_k = \nu_0$, failing the test. Note that the solution for Exercise 10.13 given by Svensén and Bishop (2009) correctly derives the result (24).

Page 489

Equation (10.107): The expectations $\mathbb{E}_{\alpha} [\ln q(\mathbf{w})]_{\mathbf{w}}$ and $\mathbb{E} [\ln q(\alpha)]$ should read $\mathbb{E}_{\mathbf{w}} [\ln q(\mathbf{w})]$ and $\mathbb{E}_{\alpha} [\ln q(\alpha)]$, respectively, where the expectation $\mathbb{E}_{\mathbf{z}}[\cdot]$ is taken over $q(\mathbf{z})$.

Page 489

Equations (10.108) through (10.112): The expectations are notationally inconsistent with (1.36); they should be of the forms shown in (10.107) or the ones corrected as above.

Page 490

The third paragraph, Line 2: A comma (,) should be inserted after the ellipsis so that the range of n should read: n = 1, ..., N.

Page 496

Equation (10.140): In order to be consistent with the mathematical notations in PRML, the differential operator d in (10.140) should be upright d. Moreover, the derivative of x with respect to x^2 should be written with parentheses as $\frac{dx}{d(x^2)}$, instead of $\frac{dx}{dx^2}$, to avoid ambiguity.

Page 501

The text after Equation (10.162): The variational parameter $\lambda(\xi)$ is a monotonic function of ξ for $\xi \geq 0$, but not that its derivative $\lambda'(\xi)$ is.

The text after Equation (10.168): A period (.) should be added at the end of the sentence that follows (10.168).

Page 512

Equation (10.222): The factor $(2\pi v_n)^{D/2}$ in the denominator of the right hand side should be omitted because it has been already included in the Gaussian in (10.213).

Page 513

Equations (10.223) and (10.224): The quantities v^{new} and \mathbf{m}^{new} in (10.223) and (10.224) are different from those in (10.217) and (10.218)⁴. Thus, we should introduce different notations, say, v and \mathbf{m} , with appropriate definitions. Specifically, one can rewrite the approximation to the model evidence in the form

$$p(\mathcal{D}) \simeq (2\pi v)^{D/2} \exp(B/2) \prod_{n=1}^{N} \left\{ s_n (2\pi v_n)^{-D/2} \right\}$$
 (25)

where

$$B = \frac{\mathbf{m}^{\mathrm{T}}\mathbf{m}}{v} - \sum_{n=1}^{N} \frac{\mathbf{m}_{n}^{\mathrm{T}}\mathbf{m}_{n}}{v_{n}}$$
 (26)

$$v^{-1} = \sum_{n=1}^{N} v_n^{-1} \tag{27}$$

$$v^{-1}\mathbf{m} = \sum_{n=1}^{N} v_n^{-1} \mathbf{m}_n. \tag{28}$$

Page 515

Equations (10.228) and (10.229): Although Svensén and Bishop (2011) correct (10.228) so that $q^{\backslash b}(\mathbf{x})$ is a normalized distribution, we do not need the normalization of $q^{\backslash b}(\mathbf{x})$ here and, even with this normalization, we cannot ensure that $\hat{p}(\mathbf{x})$ given by (10.229) is normalized. Similarly to (10.195), we can proceed with the unnormalized $q^{\backslash b}(\mathbf{x})$ given by the original (10.228) and, rather than correcting (10.228), we should correct (10.229) so that

$$\hat{p}(\mathbf{x}) \propto q^{\setminus b}(\mathbf{x}) f_b(x_2, x_3) = \dots$$
 (29)

implying that $\hat{p}(\mathbf{x})$ is a normalized distribution.

Page 515

The text after Equation (10.229): The new distribution $q^{\text{new}}(\mathbf{z})$ should read $q^{\text{new}}(\mathbf{x})$.

Page 516

Equation (10.240): The subscript k of the product $\prod_{k} \dots$ should read $k \neq j$ because we have already removed the term $\tilde{f}_{i}(\boldsymbol{\theta}_{i})$.

⁴See Svensén and Bishop (2011) for the errata for Equations (10.217) and (10.218).

Equation (11.72), Line -2: If we write the expectation $\mathbb{E}_{\mathbf{z}}[\cdot]$ taken over some given distribution $q(\mathbf{z})$ explicitly as $\mathbb{E}_{q(\mathbf{z})}[\cdot]$, the expectation in the last line but one of (11.72) should read

$$\mathbb{E}_{p_G(\mathbf{z})} \left[\exp\left(-E(\mathbf{z}) + G(\mathbf{z}) \right) \right] \tag{30}$$

where we have written the argument \mathbf{z} for $E(\mathbf{z})$ and $G(\mathbf{z})$ for clarity.

Page 556

Exercise 11.7: The interval should be $[-\pi/2, \pi/2]$ instead of [0, 1].

Page 557

Exercise 11.14, Line 2: The variance should be σ_i^2 instead of σ_i .

Page 564

The text after Equation (12.12): The derivative we consider here is that with respect to b_j (not that with respect to b_i).

Page 564

The text after Equation (12.15): The zero should be a zero vector so that we have $\mathbf{u}_i = \mathbf{0}$.

Page 575

The third paragraph, Line 5: The zero vector should be a row vector instead of a column vector so that we have $\mathbf{v}^T\mathbf{U} = \mathbf{0}^T$. Or, the both sides are transposed to give $\mathbf{U}^T\mathbf{v} = \mathbf{0}$.

Page 578

Equation (12.53): As stated in the text preceding (12.53), we should substitute $\mu = \bar{\mathbf{x}}$ into (12.53).

Page 578

The text before Equation (12.56): For the maximization with respect to \mathbf{W} , we use (C.25) and (C.27) instead of (C.24).

Page 579

Line 5: The eigendecomposition requires $O(D^3)$ computations.

Page 599

Exercise 12.1, Line -1: The quantity λ_{M+1} is an eigenvalue (not an eigenvector).

Page 602

Exercise 12.25, Line 2: The latent space distribution should read $p(\mathbf{z}) = \mathcal{N}(\mathbf{z}|\mathbf{0}, \mathbf{I})$.

The first paragraph, Line -5: The text "our predictions ..." should read: "our predictions for \mathbf{x}_{n+1} depend on all the previous observations."

Page 620

The second paragraph and the following (unlabeled) equation: The last sentence before the equation and the equation each should terminate with a period (.).

Page 621

Figures 13.12 and 13.13: It should be clarified that, similarly to $\alpha(z_{nk})$ and $\beta(z_{nk})$, the notation $p(\mathbf{x}_n|z_{nk})$ is used to denote the value of $p(\mathbf{x}_n|\mathbf{z}_n)$ when $z_{nk}=1$.

Page 622

The second paragraph, Line -1: "we see" should be omitted.

Page 623

The first paragraph, Line -2: z_{nk} should read $z_{n-1,k}$.

Page 631

Equation (13.73): The equation should read

$$\sum_{r=1}^{R} \ln \left\{ \frac{p\left(\mathbf{X}_{r} | \boldsymbol{\theta}_{m_{r}}\right) p(m_{r})}{\sum_{l=1}^{M} p\left(\mathbf{X}_{r} | \boldsymbol{\theta}_{l}\right) p(l)} \right\}. \tag{31}$$

Page 637

Equations (13.81), (13.82), and (13.83): The distribution (13.81) over \mathbf{w} should read

$$p(\mathbf{w}) = \mathcal{N}(\mathbf{w}|\mathbf{0}, \mathbf{\Gamma}) \tag{32}$$

and so on.

Page 638

The first paragraph, Line 2: "conditional on" should read "conditioned on."

Page 641

The text after Equation (13.103): The form of the Gaussian is unclear. Since a multivariate Gaussian is usually defined over a column vector, we should construct a column vector from the concerned random variables to clearly define the mean and the covariance. Specifically, the text should read for example: ..., we see that $\xi(\mathbf{z}_{n-1}, \mathbf{z}_n)$ is a Gaussian of the form

$$\xi(\mathbf{z}_{n-1}, \mathbf{z}_n) = \mathcal{N}\left(\begin{pmatrix} \mathbf{z}_{n-1} \\ \mathbf{z}_n \end{pmatrix} \middle| \begin{pmatrix} \hat{\boldsymbol{\mu}}_{n-1} \\ \hat{\boldsymbol{\mu}}_n \end{pmatrix}, \begin{pmatrix} \hat{\mathbf{V}}_{n-1} & \hat{\mathbf{V}}_{n-1,n} \\ \hat{\mathbf{V}}_{n-1,n}^{\mathrm{T}} & \hat{\mathbf{V}}_n \end{pmatrix}\right)$$
(33)

where the mean $\hat{\boldsymbol{\mu}}_n$ and the covariance $\hat{\mathbf{V}}_n$ of \mathbf{z}_n are given by (13.100) and (13.101), respectively; and the covariance $\hat{\mathbf{V}}_{n-1,n}$ between \mathbf{z}_{n-1} and \mathbf{z}_n is given by

$$\hat{\mathbf{V}}_{n-1,n} = \operatorname{cov}\left[\mathbf{z}_{n-1}, \mathbf{z}_{n}\right] = \mathbf{J}_{n-1}\hat{\mathbf{V}}_{n}.$$
(34)

Pages 642 and 643

Equation (13.109) and the following equations: If we follow the notation in Chapter 9, the typeface of the Q function should be Q.

Page 642

Equation (13.109): If we follow the notation for a conditional expectation given by (1.37), Equation (13.109) should read

$$Q\left(\boldsymbol{\theta}, \boldsymbol{\theta}^{\text{old}}\right) = \mathbb{E}_{\mathbf{Z}}\left[\ln p\left(\mathbf{X}, \mathbf{Z} | \boldsymbol{\theta}\right) \middle| \mathbf{X}, \boldsymbol{\theta}^{\text{old}}\right]$$
(35)

$$= \int d\mathbf{Z} \, p\left(\mathbf{Z} \middle| \mathbf{X}, \boldsymbol{\theta}^{\text{old}}\right) \ln p\left(\mathbf{X}, \mathbf{Z} \middle| \boldsymbol{\theta}\right)$$
(36)

which corresponds to (9.30).

Page 643

Equation (13.111): $\mathbf{V}_0^{\text{new}}$ should read $\mathbf{P}_0^{\text{new}}$. Svensén and Bishop (2011) have failed to mention (13.111).

Page 643

Equation (13.114): The size of the opening curly brace '{' should match that of the closing curly brace '}'.

Page 647

Figure 13.23, Line -1: $p(\mathbf{x}_{n+1}|\mathbf{z}_{n+1}^{(l)})$ should read $p(\mathbf{x}_{n+1}|z_{n+1}^{(l)})$.

Page 649

Exercise 13.14, Line 1: (8.67) should be (8.64).

Page 658

Figure 14.1, the equation below: The subscript of the summation in the right hand side should read m = 1.

Page 668

Equation (14.37): The arguments of the probability are notationally inconsistent with those of (14.34), (14.35), and (14.36). Specifically, the conditioning on ϕ_n should read that on t_n and the probability $p(k|\dots)$ be the value of $p(\mathbf{z}_n|\dots)$ when $z_{nk} = 1$, which we write $p(z_{nk} = 1|\dots)$. Moreover, strictly speaking, the old parameters $\pi_k, \mathbf{w}_k, \beta$ should read $\pi_k^{\text{old}}, \mathbf{w}_k^{\text{old}}, \beta^{\text{old}} \in \boldsymbol{\theta}^{\text{old}}$. In order to solve these problems, we should rewrite Equation (14.37) as, for example,

$$\gamma_{nk} = \mathbb{E}\left[z_{nk}\middle|t_n, \boldsymbol{\theta}^{\text{old}}\right] \tag{37}$$

where we have written the conditioning in the expectation explicitly and the expectation is given by

$$\mathbb{E}\left[z_{nk}|t_n,\boldsymbol{\theta}\right] = p\left(z_{nk} = 1|t_n,\boldsymbol{\theta}\right) = \frac{\pi_k \mathcal{N}\left(t_n \middle| \mathbf{w}_k^{\mathrm{T}} \boldsymbol{\phi}_n, \beta^{-1}\right)}{\sum_j \pi_j \mathcal{N}\left(t_n \middle| \mathbf{w}_j^{\mathrm{T}} \boldsymbol{\phi}_n, \beta^{-1}\right)}.$$
(38)

The unlabeled equation between (14.37) and (14.38): If we write the implicit conditioning in the expectation explicitly (similarly to the above equations), the unlabeled equation should read

$$Q\left(\boldsymbol{\theta}, \boldsymbol{\theta}^{\text{old}}\right) = \mathbb{E}_{\mathbf{Z}}\left[\ln p(\mathbf{t}, \mathbf{Z}|\boldsymbol{\theta}) \middle| \mathbf{t}, \boldsymbol{\theta}^{\text{old}}\right]$$

$$= \dots$$
(39)

$$= \dots \tag{40}$$

where we have again used the typeface Q for the Q function so that the notation is consistent with that of Chapter 9.

Page 669

Equations (14.40) and (14.41): The left hand sides should read a zero vector $\mathbf{0}$.

Page 669

Equation (14.41): Φ is undefined. The text following (14.41) should read: ... where \mathbf{R}_k $\operatorname{diag}(\gamma_{nk})$ is a diagonal matrix of size $N \times N$ and $\Phi = (\phi_1, \dots, \phi_N)^{\mathrm{T}}$ is an $N \times M$ matrix. Here, N is the size of the data set and M is the dimensionality of the feature vectors ϕ_n .

Page 669

Equation (14.43): '+const' should be added to the right hand side.

Page 671

The text after Equation (14.46): The text should read: ... where we have omitted the dependence on $\{\phi_n\}$ and defined $y_{nk} = \dots$ Or, ϕ should have been omitted from the left hand side of (14.45).

Page 671

Equation (14.48): The notation should be corrected similarly to the above erratum regarding (14.37).

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Equation (14.49): The notation should be corrected similarly to the above erratum regarding the unlabeled equation between (14.37) and (14.38).

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Equation (14.52): The negation must be removed so that $\mathbf{H}_k \equiv \nabla_k \nabla_k \mathcal{Q}$ where

$$\nabla_k \nabla_k \mathcal{Q} = -\sum_{n=1}^N \gamma_{nk} y_{nk} (1 - y_{nk}) \phi_n \phi_n^{\mathrm{T}}.$$
 (41)

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Exercise 14.1, Line 1: "of" should be inserted after "set."

Line -3: The comma in the first inline math should be removed so that the product should read: $m \times (m-1) \times \cdots \times 2 \times 1$.

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Equation (B.25): The differential operator d should be upright d.

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Line 1: 'Gamma' should read 'gamma' (without capitalization).

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Line 1: 'positive-definite' should read 'positive definite' (without hyphenation).

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Equation (B.49): \mathbf{x} in the right hand side should read \mathbf{x}_a .

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