

$$a) p(x_2) = N(\mu_1, \Sigma_{11}) = N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 6 & 8 \\ 8 & 13 \end{bmatrix}\right)$$

$$b) p(x_2) = N(\mu_2, \Sigma_{22}) = N(5, 14)$$

$$c) \text{ we know that } p(x_1 | x_2) = N(\mu_{1|2}, \Sigma_{1|2}) \quad \text{Saw soln}$$

$$\mu_{1|2} = \mu_1 + \Sigma_{12} \Sigma_{22}^{-1} (x_2 - \mu_2) = \frac{1}{14} \begin{bmatrix} 5 \\ 11 \end{bmatrix} (x_2 - 5)$$

$$\text{while } \Sigma_{1|2} = \Sigma_{11} - \Sigma_{12} \Sigma_{22}^{-1} \Sigma_{21} = \begin{bmatrix} 6 & 8 \\ 8 & 13 \end{bmatrix} - \frac{1}{14} \begin{bmatrix} 5 \\ 11 \end{bmatrix} \begin{bmatrix} 5 & 11 \end{bmatrix}$$

$$\text{Saw solution to check answer} = \begin{bmatrix} 59/14 & 57/14 \\ 57/14 & 61/14 \end{bmatrix}$$

$$d) \text{ we know that}$$

$$p(x_2 | x_1) = N(\mu_{2|1}, \Sigma_{2|1})$$

$$\mu_{2|1} = \mu_2 + \Sigma_{21} \Sigma_{11}^{-1} (x_1 - \mu_1) = 5 + \begin{bmatrix} 5 & 11 \end{bmatrix} \begin{bmatrix} 6 & 8 \\ 8 & 13 \end{bmatrix}^{-1} (x_1 - \mu_1)$$

$$= 5 + \begin{bmatrix} -\frac{23}{14} & \frac{13}{7} \end{bmatrix} x_1 - 0$$

checked soln:

$$\Sigma_{2|1} = \Sigma_{22} - \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12} = 14 - \begin{bmatrix} 5 & 11 \end{bmatrix} \begin{bmatrix} 6 & 8 \\ 8 & 13 \end{bmatrix}^{-1} \begin{bmatrix} 5 \\ 11 \end{bmatrix}$$

$$= \frac{25}{14}$$

2a) * saw soln:

$$P(y=1 | x, \theta) = \sigma(\theta^T x)$$

$$\text{with } n\ell(\theta) = -\sum_i y_i \log \sigma(\theta^T x_i) + (1-y_i) \log(1-\sigma(\theta^T x_i)) + \frac{\lambda}{2} \|\theta\|_2^2$$

$$\begin{aligned} \nabla \ell &= \sum_i y_i (1 - \sigma(\theta^T x_i)) x_i - (1 - y_i) \sigma(\theta^T x_i) x_i + \lambda \theta \\ &= \sum_i [y_i - \sigma(\theta^T x_i)] x_i + \lambda \theta \\ &= X^T (\sigma(X\theta) - y) + \lambda \theta \end{aligned}$$

* checked soln:

$$\nabla^2 \ell = \frac{d}{d\theta} \nabla \ell$$

$$= \sum_i \nabla \sigma(\theta^T x_i) x_i^T + \lambda I$$

$$= X^T \text{diag}[\sigma(X\theta)(1-\sigma(X\theta))] X + \lambda I$$

$$\begin{aligned} \text{b) } P(y=c | x, w) &= \frac{1}{Z} \exp(w_c^T x) \\ &= \frac{\exp(w_c^T x)}{\sum_i \exp(w_i^T x)} \end{aligned}$$

* saw soln: Assuming ^{each of} columns have a gaussian dist

$$n\ell(w) = -\log \prod_i \prod_c \mu_{ic}^{y_{ic}} - \lambda \text{tr}(w^T w)$$

$$= \sum_i \sum_c y_{ic} \log \mu_{ic} + \lambda \text{tr}(w^T w)$$

where

$$\nabla n\ell = X^T (N - y) + \lambda w$$