

Math Review Homework. Probability Problem Set 2.

1. Show that $\text{Bin}(m|N, p)$ is a PDF, that is, that $\sum_{m=0}^N \text{Bin}(m|N, p) = 1$.
2. Show that univariate and multivariate Gaussians are conjugate priors on the mean of univariate and multivariate Gaussians.
3. Calculate $f(\mathbf{p}|\mathbf{a})$ that is proportional to the posterior if the prior is $\text{Dir}(\mathbf{p}|(2, 4, 3))$ and the likelihood is $\text{Mult}((8, 3, 2)|\mathbf{p})$
4. Derive the log-likelihood function of a Gaussian given a set of independent samples $\mathcal{X} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N\}$
5. Derive the log-likelihood function of a Multinomial given a set of independent samples $\mathcal{X} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N\}$ where each \mathbf{x}_n is of the form

$$x_{n,m} = 1 \text{ if the outcome of experiment } n \text{ is the } m^{\text{th}} \text{ outcome} \quad (\text{B.1})$$

$$= 0 \text{ otherwise.} \quad (\text{B.2})$$

6. Suppose

$\mathcal{C} = \{\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_M\} \subset \mathbb{R}^B$ is a set of parameter vectors,

$\mathcal{X} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N\} \subset \mathbb{R}^B$ is a set of data vectors,

$\mathcal{Y} = \{y_1, y_2, \dots, y_N\} \subset \mathbb{R}$,

$\mathbf{w} \in \mathbb{R}^M$.

Define

$$K(\mathbf{x}, \mathbf{c}) = e^{-\frac{(\mathbf{x}-\mathbf{c})^t(\mathbf{x}-\mathbf{c})}{2\sigma^2}} \text{ and}$$

$$f(\mathbf{x}_n; \mathcal{C}, \mathbf{w}, \sigma) = \sum_{m=1}^M w_m K(\mathbf{x}_n, \mathbf{c}_m)$$

$$J = \sum_{n=1}^N (y_n - f(\mathbf{x}_n; \mathcal{C}, \mathbf{w}, \sigma))^2.$$

Use gradient descent to define update formulas for \mathbf{c}_m and σ to minimize J .

7. Solve the problem: Minimize $J(x, y) = 4x^2 + 9y^2$ subject to $x^2 + y^2 = 1$.