SUSTech

Machine-learning

Fall 2020

Homework 4

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4.1. Show that maximization of the class separation criterion given by $m_2 - m_1 = w^T(m_2 - m_1)$ with respect of w, using a Lagrange multiplier to enforce the constraint $w^T w = 1$, leads to the result that $w \propto (m_2 - m_1)$.

Solution. We construct the Lagrangian function

$$L = w^{T}(m_{2} - m_{1}) + \lambda(w^{T}w - 1)$$

Taking the gradient of L we obtain

$$\delta L = m_2 - m_1 + 2\lambda w$$

setting this gradient to zeros

$$w = -\frac{1}{2\lambda}(m_2 - m_1)$$

and we get

$$w \propto (m_2 - m_1)$$

4.2. Show that the Fisher criterion

$$j(w) = \frac{(m_2 - m_1)^2}{s_1^2 + s_2^2}$$

can be written in the form

$$j(w) = \frac{w^w S_B w}{w^T S_W w}$$

Solution.

$$(m_2 - m_1)^2 = (w^T (m_2 - m_1))^2$$
$$= w^T (m_2 - m_1)(m_2 - m_1)^T w$$
$$= w^T S_B w$$

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4.3. Consider a agenerative classification model for K classes defined by prior class probabilities $p(C_k) = \pi_k$ and general class-conditional dendities $p(\phi|C_k)$ where ϕ is the input feature vector. Suppose we are given a training data set $\{\phi_n, t_n\}$ where n = 1, ..., N, and t_n is a binary target vector of length K that uses the 1-of-K coding scheme, so that it has components $t_{nj} = I_{jk}$ if pattern n is from class C_k . Assuming that data points are drown independently from this model, show that the maximum-likelihood solution for the prior probabilities is given by

$$\pi_k = \frac{N_k}{N}$$

where N_k is the number of data points assigned to class C_k .

Solution. The likelihood is

$$p(\{\phi_n, t_n\} | \pi_k) = \prod_{n=1}^{N} \prod_{k=1}^{K} \{p(\phi_n | C_k) \pi_k\}^{t_{nk}}$$

taking the logarithm

$$\ln p(\{\phi_n, t_n\} | \pi_k) = \sum_{n=1}^{N} \sum_{k=1}^{K} t_{nk} \{ \ln p(\phi_n | C_k) + \ln \pi_k \}$$

with the constraint $\sum_{k} \pi_{k} = 1$, and we use Lagrange

$$\ln p(\{\phi_n, t_n\} | \{\pi_k\}) + \lambda (\sum_{k=1}^K \pi_k - 1)$$
$$\sum_{n=1}^N \frac{t_{nk}}{\pi_k} + \lambda = 0$$
$$-\pi_k \lambda = \sum_{n=1}^N t_{nk} = N_k$$

4.4. Verify the relation

$$\frac{d\sigma}{da} = \sigma(1 - \sigma)$$

for the derivative of the logistic sigmoid function defined by

$$\sigma(a) = \frac{1}{1 + \exp(-a)}$$

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Solution.

$$\frac{da}{d\sigma} = \frac{1}{\frac{d\sigma}{da}}$$

$$= \frac{1}{\sigma(1-\sigma)}$$

$$= \frac{1}{\sigma} + \frac{1}{1-\sigma}$$

$$a(\sigma) = \ln \sigma - \ln(1-\sigma)$$

$$= \ln \frac{\sigma}{1-\sigma}$$

$$e^a = \frac{\sigma}{1-\sigma}$$

$$\sigma = \frac{e^a}{1+e^a}$$

$$= \frac{1}{1+\exp(-a)}$$

4.5. By making use of the result

$$\frac{d\sigma}{da} = \sigma(1 - \sigma)$$

for the derivative of the logistic sigmoid, show that the derivative of the error function for the logistic regression model is given by

$$\nabla \mathbb{E}(w) = \sum_{n=1}^{N} (y_n - t_n) \phi_n$$

4.6. There are several possible ways in which to generalize the concept of linear discriminant functions from two classes to c classes. One possibility would be to use (c-1) linear discrimminant functions, such that $y_k(x) > 0$ for inputs x in class C_k and $y_k(x) < 0$ for not in class C_k . By drawing a simple example in two dimensions for c = 3, show that this approach can lead to regions of x-space for which the classification is ambiguous. Another approach would be to use one discriminant function $y_{jk}(x)$ for each possible pair of classes C_j and C_k , such that $y_{jk}(x) > 0$ for patterns in class C_j and $y_{jk}(x) < 0$ for patterns in class C_k . For c classes, we would need c(c-1)/2 discriminant functions. Again, by drawing a specific example in two dimensions for c = 3, show that this approach can also lead to ambiguous regions.

Solution. 内容...

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4.7. Given a set of data points $\{x^n\}$ we can define the convex hull to be the set of points x given by

$$x = \sum_{n} \alpha_n x^n$$

where $\alpha_n >= 0$ and $\sum_n \alpha_n = 1$. Consider a second set of point $\{z^m\}$ and its corresponding convex hull. The two sets of points will be linearly separable if there exists a vector \hat{w} and a scalar ω_0 such that $\hat{w}^T x^n + \omega_0 > 0$ for all x^n , and $\hat{w}^T z^m + \omega_0 < 0$ for all z^m . Show that, if their convex hulls intersect, the two sets of points cannot be linearly separable, and conversely that, if they are linearly separable, their convex hulls do not intersect.

Solution. 内容...