Solutions to Homework Set Three

ECE 271A

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1.

a) To minimize $f(\theta) = ||\mathbf{z} - \mathbf{\Phi}\theta||^2$ we compute the gradient

$$\nabla_{\theta} f = -2\mathbf{\Phi}^T (\mathbf{z} - \mathbf{\Phi}\theta)$$

and set it to zero, which leads to the standard least squares solution

$$\theta^* = (\mathbf{\Phi}^T \mathbf{\Phi})^{-1} \mathbf{\Phi}^T \mathbf{z}.$$

The matrix $(\Phi^T\Phi)^{-1}\Phi^T$ is referred to as the *pseudo-inverse* of Φ . To check that we have a minimum we compute the Hessian

$$\nabla_{\theta}^{2} f = 2 \left(\mathbf{\Phi}^{T} \mathbf{\Phi} \right)^{T} = 2 \mathbf{\Phi}^{T} \mathbf{\Phi}$$

and use the theorem, which we already mentioned in the last homework set, that \mathbf{A} is positive definite if and only if there is a non-singular matrix \mathbf{R} such that $\mathbf{A} = \mathbf{R}^T \mathbf{R}$. (See e.g. *Linear Algebra and its Applications* by G. Strang, Harcourt Brace Jovanovic, 1988.) Since this is exactly the case of our Hessian, we only have to prove that $\mathbf{\Phi}$ is non-singular. We cannot really do this without knowing what the sample points x_i are but, assuming that all the x_i are different, this will hold.

b) Since, when x is known, $f(x,\theta)$ is a deterministic function of θ , and $\epsilon \sim \mathcal{N}(0,\sigma^2)$ it follows that

$$P_{Z|X}(z|x;\theta) = \mathcal{G}(x, f(x, \theta), \sigma^2)$$

Given an iid sample $\mathcal{D} = \{\mathcal{D}_x, \mathcal{D}_y\}$, the ML estimate of θ is then

$$\theta^* = \arg \max_{\theta} \sum_{i} \log P_{Z|X}(z_i|x_i;\theta)$$

$$= \arg \max_{\theta} \sum_{i} -\frac{(z_i - f(x_i,\theta))^2}{2\sigma^2} - \frac{n}{2} \log(2\pi\sigma^2)$$

$$= \arg \min_{\theta} \sum_{i} (z_i - f(x_i,\theta))^2$$

$$= \arg \min_{\theta} \sum_{i} (z_i - \phi_i\theta))^2$$
(1)

where $\phi_i=(1,x_i,\dots,x_i^K)$, i.e. the i^{th} row of Φ . It follows that the ML solution is the one that minimizes $||\mathbf{z}-\Phi\theta||^2$ and therefore the same as the least squares solution.

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