

LAB 3 THEORY

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1. PROBLEM 1:

1.1. Problem Statement. Derive a formula for which the coefficients $a^0, a^1, a^2, \dots, a^m$, for which the linear function $y(x^1, x^2, \dots, x^m) = a^0 + a^1x^1 + a^2x^2 + \dots + a^mx^m$ minimizes

$$E_2(a^0, a^1, a^2, \dots, a^m) = \sum_{i=1}^N (y_i - (a^0 + a^1x_i^1 + a^2x_i^2 + \dots + a^mx_i^m))^2.$$

1.2. Derivation. Consider $y(x^1, x^2, \dots, x^m)$ as a sum of $m + 1$ functions, each of the form $\phi_i = x^i$ defining $\phi_0 = 1$. Then we have $y(\mathbf{x}, \mathbf{a}) = \mathbf{a}^T \phi(\mathbf{x})$. Now we must find the parameters \mathbf{a}^T .

We are given the cost function E_2 , so to find a minimum, we set the gradient of the cost function to zero, that is $\nabla_{\mathbf{a}} E_2(\mathbf{a}) = 0$. For this cost function, this solves to $0 = \sum_{i=1}^N y_i \phi(\mathbf{x}_i)^T - \mathbf{a}^T (\sum_{i=1}^N \phi(\mathbf{x}_i) \phi(\mathbf{x}_i)^T)$.

This is then solved for the coefficients \mathbf{a} and translated into matrix notation, giving

$$(1) \quad \mathbf{a} = (\Phi^T \Phi)^{-1} \Phi^T \mathbf{y}.$$

Where $\mathbf{y} = (y_1, \dots, y_N)^T$ and Φ is the matrix of predictor values where each row is given by $\Phi(i, :) = (1, x_i^1, \dots, x_i^m)$.

In practice, this is simpler to solve in the form

$$(2) \quad A\mathbf{a} = \Phi^T \mathbf{y},$$

where $A = \Phi^T \Phi$.

Because of the structure of Φ , this results in

$$(3) \quad (A)_{ij} = \sum_{l=1}^N x_l^i x_l^j.$$

Because superscripts here are *not* exponents, this is not x_l^{i+j} . Similarly, the right hand side solves to be

$$(4) \quad b_i = \sum_{l=1}^N y_l x_l^i.$$

This pair of equations is what must be solved to find the coefficients \mathbf{a} .