

Affine Mmse Predictors

We want to find A and b to minimize

$$\mathsf{E} |Ax + b - y|^2.$$

Proof. We can express $\mathsf{E}(|Ax+b-y|^2)$ as $\mathsf{E}((Ax+b-y)^\top(Ax+b-y))$

The gradients with respect to b are

$$+ 0 + A E(x) - 0$$

 $+ A E(x) + 2b - E(y)$
 $- 0 - E(y) + 0$

so $2A \mathbf{E}(x) + 2b - 2\mathbf{E}(y)$. The gradients with respect to A are

so $2 \mathsf{E}(xx^\top)A^\top + 2 \mathsf{E}(x)b^\top - 2 \mathsf{E}(xy^\top)$. We want A and b solutions to

$$A \mathbf{E}(x) + b - \mathbf{E}(y) = 0$$

$$\mathbf{E}(xx^{\top})A^{\top} + \mathbf{E}(x)b^{\top} - \mathbf{E}(xy^{\top}) = 0$$

so first get $b = \mathsf{E}(y) - A\,\mathsf{E}(x)$. Then express

$$\begin{split} \mathbf{E}(xx^\top)A^\top + \mathbf{E}(x)(\mathbf{E}(y) - A\,\mathbf{E}(x))^\top - \mathbf{E}(xy^\top) &= 0. \\ \mathbf{E}(xx^\top)A^\top + \mathbf{E}(x)\,\mathbf{E}(y)^\top - \mathbf{E}(x)\,\mathbf{E}(x)^\top A^\top - \mathbf{E}(xy^\top) &= 0. \\ (\mathbf{E}(xx^\top) - \mathbf{E}(x)\,\mathbf{E}(x)^\top)A^\top &= \mathbf{E}(xy^\top) - \mathbf{E}(x)\,\mathbf{E}(y)^\top. \\ \mathbf{cov}(x,x)A^\top &= \mathbf{cov}(x,y). \end{split}$$

So $A^{\top} = \mathbf{cov}(x,x)^{-1}\mathbf{cov}(x,y)$ means $A = \mathbf{cov}(y,x)\mathbf{cov}(x,x)^{-1}$ is a solution. Then $b = \mathbf{E}(y) - \mathbf{cov}(y,x)\mathbf{cov}(x,x)^{-1}\mathbf{E}(x)$. So to summarize, the estimator $\phi(x) = Ax + b$ is

$$\operatorname{cov}(y,x)\left(\operatorname{cov} x,x\right)^{-1}x+\operatorname{E}(y)-\operatorname{cov}(y,x)\operatorname{cov}(x,x)^{-1}\operatorname{E}(x)$$

or

$$\mathsf{E}(y) + \mathsf{cov}(y, x) \left(\mathsf{cov}\, x, x\right)^{-1} \left(x - \mathsf{E}(x)\right)$$