Solutions to selected exercises from Chapter 21 of Wasserman — All of Statistics

(1) We have  $R(J) = \int b(x)^2 dx + \int v(x) dx$ . Furthermore,

$$b(x) = \mathbb{E}(\hat{f}(x)) - f(x) = \sum_{j=1}^{J} \mathbb{E}(\hat{\beta}_j)\phi_j(x) - f(x) = \sum_{j=1}^{J} \beta_j \phi_j(x) - f(x) = \sum_{j=J+1}^{\infty} \beta_j \phi_j(x).$$

By orthogonality of the  $\phi_i$ ,

$$\int b^2 = \sum_{j=J+1}^{\infty} \beta_j^2$$

For the other term,

$$v(x) = \mathbb{E}\left(\hat{f}(x) - \mathbb{E}\hat{f}(x)\right)^2 = \mathbb{E}\left(\sum_{j=1}^J (\hat{\beta}_j - \beta_j)\phi_j(x)\right)^2 = \sum_{j=1}^J \sum_{k=1}^J \operatorname{Cov}(\hat{\beta}_j, \hat{\beta}_k)\phi_j(x)\phi_k(x).$$

Again using orthogonality of the  $\phi_j$  yields

$$\int v(x)dx = \sum_{j=1}^{J} \mathbb{V}(\hat{\beta}_j) = \sum_{j=1}^{J} \frac{\sigma_j^2}{n}.$$

Adding  $\int b^2$  to  $\int v$  proves the theorem.

(4) Set  $f_n = \sum_{j=1}^n \beta_j \phi_j$ . By hypothesis,  $\int (f - f_n)^2 \to 0$  as  $n \to \infty$ . We have

$$\int (f - f_n)^2 = \int f^2 - 2 \int f f_n + \int f_n^2.$$

By the Cauchy-Schwarz inequality (equivalently Hölder's inequality),

$$\left| \int f f_n \right| \le \sqrt{\int f^2} \sqrt{\int f_n^2}.$$

Hence,

$$\int (f - f_n)^2 \ge \int f^2 - 2\sqrt{\int f^2} \sqrt{\int f_n^2} + \int f_n^2 = \left(\sqrt{\int f^2} - \sqrt{\int f_n^2}\right)^2.$$

Thus  $\int f_n^2 \to \int f^2$ . On the other hand,

$$\int f_n^2 = \sum_{j=1}^n \beta_j^2$$

by the orthogonality of the  $\phi_j$ . The proof of Hölder's inequality follows from Young's inequality for products. See Wikipedia.

- (5) See the Jupyter Notebook 5.ipynb.
- (6) See the Jupyter Notebook 6.ipynb.
- (7) See the Jupyter Notebook 9.ipynb.

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