## Chapter VII

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### Summary

1 Ex. 7.3

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Let  $\hat{\mathbf{f}} = \mathbf{S}\mathbf{y}$  be a linear smoothing of  $\mathbf{y}$ .

(a) If  $S_{ii}$  is the *i*th diagonal element of **S**, showing that for **S** arising from least squares projections and cubic smoothing splines, the cross-validated residual can be written by as

$$y_i - \hat{f}^{-i}(x_i) = \frac{y_i - \hat{f}(x_i)}{1 - S_{ii}}$$
 (1)

- **(b)** Use this result to show that  $|y \hat{f}^{-i}(x_i)| \ge |y_i \hat{f}(x_i)|$ .
- (c) Find general conditions on any smoother  ${\bf S}$  to make result (7.64) hold.

**Proof:** (a) Denote  $A = X^T X$ ; for the least square with the  $i^{th}$  data points omitted:

$$\hat{f}^{-i}(x_i) = x_i^T ((X^{-i})^T X^{-i})^{-1} (X^{-i})^T y^{-i}$$
(2)

It is true that  $X^TX = \sum_{j=1}^{j=n} x_i x_i^T$ , and  $(X^{-i})^T X^{-i} = X^T X - x_i x$ . And it is obvious that  $(X^{-i})^T y^{-i} = X^T y - x_i y_i$ . Finally, (2) can be simplified as following:

$$\hat{f}^{-i}(x_i) = x_i^T (A - x_i x_i^T)^{-1} (X^T y - x_i y_i)$$
(3)

There exists a tricks when simplifying (3). In mathematics, the *Sherman-Morrison formula* (see https://en.wikipedia.org/wiki/Sherman-Morrison\_formula) computes the inverse of the sun of an invertible matrix A and the outer product,  $uv^T$ :

$$(A - x_i x_i^T)^{-1} = A^{-1} + \frac{A^{-1} x_i x_i^T A^{-1}}{1 - x_i^T A^{-1} x_i}$$
(4)

Since  $X^T \in \mathfrak{R}^{pn}$  equals to  $[x_1, ..., x_i, ..., x_n]$ , we can write  $x_i^T$  as  $\delta_i^T X(\delta_i$  is a vector  $\delta \in \mathfrak{R}^n$  where all elements expect the  $i^{th}$  component valued as 1 is zero, timely  $\delta_i^T = [0, ..., 1^i, ..., 0]$ ). We denote:

$$x_i^T A^{-1} x_i = \delta_i^T X A^{-1} X^T \delta_i \tag{5}$$

$$= \delta_i^T S \delta_i \tag{6}$$

$$=S_{ii} \tag{7}$$

where  $S = X(X^TX)^{-1}X^T$ . To simplify (3), we substitute (4) (7) into this equation and we get:

$$\hat{f}^{-i}(x_i) = x_i^T \left[ A^{-1} + \frac{A^{-1} x_i x_i^T A^{-1}}{1 - S_{ii}} \right]$$

$$= x_i^T A^{-1} X^T y - x_i^T A^{-1} x_i y_i + \frac{x_i^T A^{-1} x_i (x_i^T A^{-1} X^T y) - (x_i^T A^{-1} x_i)^2 y_i}{1 - S_{ii}}$$
(9)

$$=\hat{f}(x_i) - S_{ii}y_i + \frac{S_{ii}\hat{f}(x_i) - S_{ii}^2y_i}{1 - S_{ii}}$$
(10)

$$=\frac{\hat{f}(x_i) - y_i S_{ii} - \hat{f}(x_i) S_{ii} + y_i S_{ii}^2 + \hat{f}(x_i) S_{ii} - y_i S_{ii}^2}{1 - S_{ii}}$$
(11)

$$=\frac{\hat{f}(x_i) - y_i S_{ii}}{1 - S_{ii}} \tag{12}$$

$$=\frac{\hat{f}(x_i) - y_i + y_i(1 - S_{ii})}{1 - S_{ii}}$$
(13)

$$= y_i - \frac{y_i - \hat{f}(x_i)}{1 - S_{ii}} \tag{14}$$

## List of Listings

Adding the comma-separated parameter caption=Python example inside the brackets, enables the caption. This caption can be later used in the list of Listings.

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# Listings