Appendix A

Evolution of the expected value of $S^2(t)$ after a leap of the mean value

Let us consider a stochastic process whose parameters μ and σ change suddenly in a concrete moment. May $Y = X(\mu_1, \sigma_1)$ and $Z = X(\mu_2, \sigma_2)$ be the distributions of the process before and after the leap respectively. The variable x_i can be defined by

$$x_i = \begin{cases} y_i, & \text{if } i \leq a; \\ z_i, & \text{if } i > a. \end{cases}$$
 (A.1)

where a is the last sample before the leap.

Let us consider a sample $x = \{x_1, \ldots, x_n\}$ of size N. The expression of his mean value \bar{x} can be written as follows:

$$\bar{x} = \frac{a}{N}\bar{y} + \frac{(N-a)}{N}\bar{z} \tag{A.2}$$

Parameter a indicates the number of elements x_i in x corresponding to the the first distribution Y. As a takes values from N to 0, the expected value of \bar{x} evolves linearly from μ_1 to μ_2 .

Since in this case we are *not* dealing with a stationary process, the expected value of the *sample variance* is no longer the variance of the population $(E(S^2) \neq \sigma^2)$. The new expected value is to be calculated here depending on a as a parameter. As known, the expression of the sample variance is

$$S_a^2 \doteq \frac{1}{N-1} \sum_{i=1}^{N} (x_i - \bar{x})^2$$
 (A.3)

Considering (A.1) we can write:

$$S_a^2 = \frac{1}{N-1} \left(\sum_{i=1}^a (y_i - \bar{x})^2 + \sum_{i=a+1}^N (z_i - \bar{x})^2 \right)$$
 (A.4)

Adding and deducting μ in both addends and expanding the squares:

$$\sum_{i=1}^{a} (y_i - \bar{x})^2 = \sum_{i=1}^{a} ((y_i - \mu_1) - (\bar{x} - \mu_1))^2 =$$

$$= \sum_{i=1}^{a} ((y_i - \mu_1)^2 + (\bar{x} - \mu_1)^2) - 2 \sum_{i=1}^{a} ((y_i - \mu_1)(\bar{x} - \mu_1))$$

$$\sum_{i=a+1}^{N} (z_i - \bar{x})^2 = \sum_{i=a+1}^{N} ((z_i - \mu_2) - (\bar{x} - \mu_2))^2 =$$

$$= \sum_{i=a+1}^{N} ((z_i - \mu_2)^2 + (\bar{x} - \mu_2)^2) - 2 \sum_{i=a+1}^{N} ((z_i - \mu_2)(\bar{x} - \mu_2))$$
(A.6)

In order to find the expected value of (A.4), the expected value of the terms in (A.5) and (A.6) are calculated. For equation (A.5):

$$E\left[\left(y_i - \mu_1\right)^2\right] = \sigma_1^2 \tag{A.7}$$

Equation (A.2) is used now for the expansion of the next two terms

$$E\left[\left(\bar{x} - \mu_{1}\right)^{2}\right] =$$

$$= E\left[\left(\frac{a}{N} \cdot \bar{y} + \frac{(N-a)}{N} \cdot \bar{z} - \mu_{1}\right)^{2}\right] = E\left[\left(\frac{a}{N}(\bar{y} - \mu_{1}) + \frac{(N-a)}{N}(\bar{z} - \mu_{1})\right)^{2}\right] =$$

$$= \frac{a^{2}}{N^{2}}E\left[\left(\bar{y} - \mu_{1}\right)^{2}\right] + \frac{(N-a)^{2}}{N^{2}}E\left[\left(\bar{z} - \mu_{1}\right)^{2}\right] + 2\frac{a(N-a)}{N^{2}}E\left[\left(\bar{y} - \mu_{1}\right)(\bar{z} - \mu_{1})\right] =$$

$$= \frac{a^{2}}{N^{2}}\frac{\sigma_{1}^{2}}{a} + \frac{(N-a)^{2}}{N^{2}}E\left[\left((\bar{z} - \mu_{2}) + (\mu_{2} - \mu_{1})\right)^{2}\right] + 0 =$$

$$= \frac{a^{2}}{N^{2}}\frac{\sigma_{1}^{2}}{a} + \frac{(N-a)^{2}}{N^{2}}E\left[\left(\bar{z} - \mu_{2}\right)^{2} + (\mu_{2} - \mu_{1})^{2} + 2(\bar{z} - \mu_{2})(\mu_{2} - \mu_{1})\right] =$$

$$= \frac{a}{N^{2}}\sigma_{1}^{2} + \frac{(N-a)^{2}}{N^{2}}\left(\frac{\sigma_{2}^{2}}{(N-a)} + \Delta\mu^{2} + 0\right) =$$

$$= \frac{a}{N^{2}}\sigma_{1}^{2} + \frac{(N-a)}{N^{2}}\sigma_{2}^{2} + \frac{(N-a)^{2}}{N^{2}}\Delta\mu^{2}$$
(A.8)

$$E\left[\sum_{i=1}^{a} ((y_{i} - \mu_{1})(\bar{x} - \mu_{1}))\right] =$$

$$= a \cdot E\left[(\bar{y} - \mu_{1})(\bar{x} - \mu_{1})\right] = a \cdot E\left[(\bar{y} - \mu_{1})\left(\frac{a}{N}(\bar{y} - \mu_{1}) + \frac{(N - a)}{N}(\bar{z} - \mu_{1})\right)\right] =$$

$$= \frac{a}{N}E\left[a(\bar{y} - \mu_{1})^{2} + (N - a)(\bar{y} - \mu_{1})(\bar{z} - \mu_{1})\right] = \frac{a}{N}\left(\sigma_{1}^{2} + 0\right) =$$

$$= \frac{a}{N}\sigma_{1}^{2}$$
(A.9)

Doing the same with terms on equation (A.6):

$$E\left[(z_i - \mu_2)^2\right] = \sigma_2^2 \tag{A.10}$$

$$E\left[\left(\bar{x} - \mu_2\right)^2\right] = \frac{a}{N^2}\sigma_1^2 + \frac{(N-a)}{N^2}\sigma_2^2 + \frac{a}{N^2}\Delta\mu^2 \tag{A.11}$$

$$E\left[\sum_{i=a+1}^{N} ((z_i - \mu_2)(\bar{x} - \mu_2))\right] = \frac{(N-a)}{N} \sigma_2^2$$
 (A.12)

Considering equations (A.7), (A.8), (A.9), (A.10), (A.11) and (A.12) the expected value of S_a^2 results:

$$E\left[S_a^2\right] = \frac{a}{N}\sigma_1^2 + \frac{(N-a)}{N}\sigma_2^2 + \frac{a(N-a)}{N(N-1)}\Delta\mu^2$$
 (A.13)