

Covariance matrix for histogram made using seed events

Consider a histogram created by generating N_{seed} “seed events” according to a given distribution, and then using each seed event to generate a further number N_{sim} events by smearing with a certain resolution function, which results in a final histogram with n_i entries in the i th bin. Because each seed event contributes N_{sim} times to the histogram, there are correlations between bins. This note gives an expression for the covariance $\text{cov}[n_i, n_j]$ between the number of entries in any pair of bins. The result is given in equation (12) below for the case where the number of seed events N_{seed} is treated as a constant and by equation (18) for the case where N_{seed} is a Poisson variable with mean ν_{seed} .

The result for $\text{cov}[n_i, n_j]$ of course also provides the standard deviations

$$\sigma[n_i] = \sqrt{\text{cov}[n_i, n_i]}, \quad (1)$$

and the matrix of correlation coefficients,

$$\rho_{ij} = \frac{\text{cov}[n_i, n_j]}{\sigma[n_i]\sigma[n_j]}. \quad (2)$$

1 Case of fixed number of seed events

In this section we consider the case where the number of seed events, N_{seed} , is taken as a constant; it does not fluctuate upon repetition of the experiment. Suppose the indices i, j , denote a bin of the final histogram (after smearing) and let s be a bin in which the seed event is found. The seed events are labelled with indices $a, b = 1, \dots, N_{\text{seed}}$. Let n_{ia} be the number of events found in bin i from seed event a . The total number of entries found in bin i is obtained by summing over all seed events:

$$n_i = \sum_{a=1}^{N_{\text{seed}}} n_{ia}. \quad (3)$$

The covariance $\text{cov}[n_i, n_j]$ is given by

$$\text{cov}[n_i, n_j] = E[n_i n_j] - E[n_i]E[n_j]. \quad (4)$$

We can find $E[n_i]$ by first considering the case of a single seed event a . This is

$$\begin{aligned}
E[n_{ia}] &= \sum_{n_{ia}} n_{ia} P(n_{ia}) \\
&= \sum_s \sum_{n_{ia}} n_{ia} P(n_{ia}|s) Q_s \\
&= \sum_s E[n_{ia}|s] Q_s \\
&= N_{\text{sim}} \sum_s P_{is} Q_s \equiv N_{\text{sim}} \bar{P}_i, \tag{5}
\end{aligned}$$

where the sums are over all possible values of n_{ia} (0 to N_{sim}) and s (over all bins). Here $P(n_{ia})$ is the probability to find n_{ia} entries in bin i from seed event a , $P(n_{ia}|s)$ is the corresponding conditional probability given that the seed event is in bin s , and Q_s is the probability for the seed event to be in bin s . In the third line of (5), $E[n_{ia}|s] = N_{\text{sim}} P_{is}$ is the expected number of entries in bin i given that the seed event is in bin s , and P_{is} is the probability to observe an event in bin i given that the seed event is in bin s . By symmetry this is the same for all seed events and therefore it does not depend on the index a . In the last line of (5) we used the notation

$$\bar{P}_i \equiv \sum_s P_{is} Q_s. \tag{6}$$

Summing over N_{seed} independent seed events gives

$$E[n_i] = N_{\text{seed}} N_{\text{sim}} \bar{P}_i. \tag{7}$$

We now need the expectation value $E[n_i n_j]$. This can be written

$$E[n_i n_j] = E \left[\left(\sum_{a=1}^{N_{\text{seed}}} n_{ia} \right) \left(\sum_{b=1}^{N_{\text{seed}}} n_{jb} \right) \right] = \sum_{a,b=1}^{N_{\text{seed}}} E[n_{ia} n_{jb}]. \tag{8}$$

Of the N_{seed}^2 terms in the double sum, N_{seed} have $a = b$. For the $N_{\text{seed}}^2 - N_{\text{seed}}$ terms with $a \neq b$, the seed events are independent and therefore

$$E[n_{ia} n_{jb}] = E[n_{ia}] E[n_{jb}] = N_{\text{sim}}^2 \bar{P}_i \bar{P}_j \quad (a \neq b). \tag{9}$$

For the N_{seed} terms with $a = b$, consider again a single seed event a found in a given bin s . For a fixed s , n_{ia} and n_{ja} are multinomially distributed, and therefore the conditional expectation value of $n_{ia} n_{ja}$ for fixed s is

$$E[n_{ia} n_{ja} | s] = N_{\text{sim}}^2 P_{is} P_{js} + N_{\text{sim}} P_{is} (\delta_{ij} - P_{js}). \tag{10}$$

(If one regards N_{sim} as a Poisson variable rather than fixed, then the second term, proportional to N_{sim} , in (10) is absent.) We need to average the expectation value (10) over s and multiply by N_{seed} to obtain

$$\begin{aligned}
E[n_i n_j] &= N_{\text{seed}} \sum_s E[n_{ia} n_{ja} | s] Q_s \\
&= N_{\text{seed}} \left(\sum_s N_{\text{sim}}^2 P_{is} P_{js} Q_s + \sum_s N_{\text{sim}} P_{is} (\delta_{ij} - P_{js}) Q_s \right) \\
&= N_{\text{seed}} N_{\text{sim}}^2 \overline{P_i P_j} + N_{\text{seed}} N_{\text{sim}} (\delta_{ij} \overline{P_i} - \overline{P_i P_j}) .
\end{aligned} \tag{11}$$

Putting together the ingredients gives the covariance for n_i and n_j ,

$$\text{cov}[n_i, n_j] = N_{\text{seed}} N_{\text{sim}}^2 (\overline{P_i P_j} - \overline{P_i} \overline{P_j}) + N_{\text{seed}} N_{\text{sim}} (\overline{P_i} \delta_{ij} - \overline{P_i P_j}) . \tag{12}$$

As mentioned above, if one takes N_{sim} to be Poisson distributed rather than fixed, then the second term in (12) proportional to N_{sim} is absent. The required ingredients are thus the matrix of probabilities P_{is} (probability to observe the event in bin i given a seed event in bin s), and the probability to have a seed event in bin s , Q_s , both of which can be estimated, e.g., from Monte Carlo.

2 Case of random number of seed events

In the previous section, the number of seed events N_{seed} was treated as a constant. We can also treat it as a random variable following a Poisson distribution with a mean ν_{seed} . To find the covariance $\text{cov}[n_i, n_j]$ we need the expectation values $E[n_i]$ and $E[n_i n_j]$. For $E[n_i]$ we have

$$\begin{aligned}
E[n_i] &= \sum_{N_{\text{seed}}=0}^{\infty} P(N_{\text{seed}}; \nu_{\text{seed}}) E[n_i | N_{\text{seed}}] \\
&= \sum_{N_{\text{seed}}=0}^{\infty} P(N_{\text{seed}}; \nu_{\text{seed}}) N_{\text{seed}} N_{\text{sim}} \overline{P_i} \\
&= \nu_{\text{seed}} N_{\text{sim}} \overline{P_i} ,
\end{aligned} \tag{13}$$

where $P(N_{\text{seed}}; \nu_{\text{seed}})$ is the Poisson probability for N_{seed} with mean value ν_{seed} . Equation (7) for $E[n_i]$ is with constant N_{seed} and so this was used for $E[n_i | N_{\text{seed}}]$ to obtain the second line of (13) above.

For $E[n_i n_j]$ we have

$$\begin{aligned}
E[n_i n_j] &= \sum_{N_{\text{seed}}=0}^{\infty} P(N_{\text{seed}}; \nu_{\text{seed}}) E[n_i n_j | N_{\text{seed}}] \\
&= \sum_{N_{\text{seed}}=0}^{\infty} P(N_{\text{seed}}; \nu_{\text{seed}}) \left(\sum_{a,b=1}^{N_{\text{seed}}} E[n_{ia} n_{jb} | N_{\text{seed}}] \right) .
\end{aligned} \tag{14}$$

In the sums over a and b in equation (14) there are N_{seed} terms with $a = b$, and for these we can use equation (11). For the remaining $N_{\text{seed}}(N_{\text{seed}} - 1)$ terms with $a \neq b$ we have

$$E[n_{ia}n_{jb}|N_{\text{seed}}] = E[n_{ia}|N_{\text{seed}}]E[n_{jb}|N_{\text{seed}}] = N_{\text{sim}}^2 \overline{P_i P_j} \quad (a \neq b), \quad (15)$$

because two distinct seed events are not correlated. The required expectation value for a given N_{seed} is therefore

$$\begin{aligned} E[n_i n_j | N_{\text{seed}}] &= N_{\text{seed}} E[n_{ia} n_{jb} | N_{\text{seed}}] + N_{\text{seed}}(N_{\text{seed}} - 1) E[n_{ia} | N_{\text{seed}}] E[n_{jb} | N_{\text{seed}}] \\ &= N_{\text{seed}} \left[N_{\text{sim}}^2 \overline{P_i P_j} + N_{\text{sim}}(\overline{P_i} \delta_{ij} - \overline{P_i P_j}) \right] + N_{\text{seed}}(N_{\text{seed}} - 1) N_{\text{sim}}^2 \overline{P_i P_j}. \end{aligned} \quad (16)$$

We can then use (16) together with (14). To evaluate the result we need the Poisson expectation value $E[N_{\text{seed}}]$, and we can use the Poisson variance $V[N_{\text{seed}}] = E[N_{\text{seed}}^2] - (E[N_{\text{seed}}])^2 = \nu_{\text{seed}}$ to find

$$E[N_{\text{seed}}^2] = \nu_{\text{seed}}(\nu_{\text{seed}} + 1). \quad (17)$$

Using the resulting value of $E[n_i n_j]$ together with $E[n_i]$ gives the final expression for the covariance,

$$\text{cov}[n_i, n_j] = \nu_{\text{seed}} \left[N_{\text{sim}}^2 \overline{P_i P_j} + N_{\text{sim}}(\overline{P_i} \delta_{ij} - \overline{P_i P_j}) \right]. \quad (18)$$