

Let A, B be nxm matrices, then:

1.  $C = A \odot B$  is the elementwise (Hadamard) product, i.e  $c_{ij} = a_{ij}b_{ij}$
2. Similarly,  $D = A \oslash B$  is the elementwise division:  $d_{ij} = a_{ij}/b_{ij}$
3.  $A \star B$  is the discrete cross-correlation of A and B

Further let:  $\mathbf{i} = (i, j)$  be a multi-index and  $\mathbf{h} = (h_x, h_y)$  be a two-dimensional offset.

**Starting from the sum:**

$$\begin{aligned} T &= \sum_{ij} (z_{ij}^2 - z_{ij})^2 = \sum_{h \in S_h} (z_i - z_{i+h})^2 = \sum_{h \in S_h} z_i^2 - 2z_i z_{i+h} + z_{i+h}^2 \\ &= \sum_{h \in S_h} z_i^2 - 2 \sum_{h \in S_h} z_i z_{i+h} + \sum_{h \in S_h} z_{i+h}^2 \\ &= (Z \odot Z) \star J - 2 Z \star Z + J \star (Z \odot Z) \end{aligned}$$

Where J is an all-ones matrix, with the same shape as Z

To obtain the normalizing factor (how many pairs from Z participate per lag) we can calculate:

$$D = J \star J$$

Finally, the discrete 2D correlation function / variogram is given by:

$$C = T \oslash D = [(Z \odot Z) \star J - 2 Z \star Z + J \star (Z \odot Z)] \oslash [J \star J]$$

Note: Computationally, flipping  $J \star (Z \odot Z)$  along both axes is equivalent to computing  $(Z \odot Z) \star J$ .

To obtain the distances (lags) for which C, we need to note that, if Z:  $n \times m$ , then

$Z \star Z$ :  $2n - 1 \times 2m - 1$  matrix. The matrix of lags Q for which C was computed is then given:

$$Q_{h_y, h_x} = \sqrt{(h_x \cdot \Delta x)^2 + (h_y \cdot \Delta y)^2}$$

Where:

$$h_y \in \{-(n_y - 1), -(n_y - 2), \dots, -1, 0, 1, \dots, n_y - 2, n_y - 1\} \text{ and}$$

$$h_x \in \{-(n_x - 1), -(n_x - 2), \dots, -1, 0, 1, \dots, n_x - 2, n_x - 1\}$$

**Finally:**

1. We flatten **C** and **Q**
2. We bin the data with equal population bins based on C
3. We average the data in each bin
4. Plot the bin centers vs averaged C-values