

# ch\_xcorr.m

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This document briefly explains the workings of the `ch_xcorr` algorithm.

## Algorithm

```
1 function [ccg,ic] = ch_xcorr(hc_L,hc_R,frameCount,frame_length,noverlap,maxlag,tau,  
2     ...  
3     varargin)
```

```
1 % SYNTAX  
2 %  
3 % CCG = CH_XCORR(HC_L,HC_R,FRAME_LENGTH,NOVERLAP,MAXLAG,TAU)  
4 % CCG = CH_XCORR(...,INHIB)  
5 % CCG = CH_XCORR(...,IC_T,NORM_FLAG)  
6 % CCG = CH_XCORR(...,INHIB_MODE)  
7 % [CCG,IC] = CH_XCORR(...)
```

For the purposes of this document:

$HC\_L = \mathbf{h}_L$

$HC\_R = \mathbf{h}_R$

$FRAME\_LENGTH = M$

$NOVERLAP = N$

$TAU = \tau$

$INHIB = \iota$

$IC\_T = \Theta_\chi$

$FRAME\_COUNT = \text{floor}((\max(\text{size}(HC\_L)) - MAXLAG - 1) / (FRAME\_LENGTH)) - NOVERLAP + 1 = L$

The algorithm takes two matrices of data `hc_L` and `hc_R` (such as the output of a peripheral ear model) and divides it into frames of length `frame_length` (in samples). From these data, cross-correlograms (averaged cross-correlations) are calculated for each frame. The cross-correlations and cross-correlograms can be calculated in numerous ways, with or without normalisation, and with optional inhibition.

Specifically, cross-correlations are calculated in frequency channel index  $i$ , sample index  $n$  and lag index  $m$  using the input data in the following way:

$$\hat{\mathbf{c}}(i, m, n) = \frac{\dot{\mathbf{c}}(i, m, n)}{\sqrt{\mathbf{a}_L(i, m, n)\mathbf{a}_R(i, m, n)}} \quad (1)$$

where

$$\dot{\mathbf{c}}(i, m, n) = \frac{1}{\tau} \mathbf{h}_L(i, \max(n + m, n)) \mathbf{h}_R(i, \max(n - m, n)) + \left(1 - \frac{1}{\tau}\right) \dot{\mathbf{c}}(i, m, n - 1), \quad (2)$$

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$$\mathbf{a}_L(i, m, n) = \frac{1}{\tau} \mathbf{h}_L^2(i, \max(n + m, n)) + \left(1 - \frac{1}{\tau}\right) \mathbf{a}_L(i, m, n - 1), \quad (3)$$

$$\mathbf{a}_R(i, m, n) = \frac{1}{\tau} \mathbf{h}_R^2(i, \max(n - m, n)) + \left(1 - \frac{1}{\tau}\right) \mathbf{a}_R(i, m, n - 1) \quad (4)$$

and  $\tau$  is the time constant of the exponentially decaying window in samples. This is used to extract the Interaural Coherence (IC)  $\chi$  as:

$$\chi(i, n) = \max_m \hat{\mathbf{c}}(i, m, n) \quad (5)$$

However, the data used in subsequent stages of the algorithm depends on `norm_flag`. Subsequent stages may use the normalised cross-correlation  $\hat{\mathbf{c}}$  (if `norm_flag = 1`), or the un-normalised cross-correlation  $\mathbf{c}$  (if `norm_flag  $\neq$  1`).

Following this, inhibition is applied, although this procedure depends upon the `inhib_mode` supplied to the function. The default (`'multiply'`) is a multiplication mechanism:

$$\mathbf{c}(i, m, n) = \iota(i, n) \hat{\mathbf{c}}(i, m, n) \quad (\text{norm\_flag} \neq 1) \quad (6)$$

or

$$\mathbf{c}(i, m, n) = \iota(i, n) \hat{\mathbf{c}}(i, m, n) \quad (\text{norm\_flag} = 1) \quad (7)$$

Alternatively, a subtractive procedure (`'subtract'`) may be used:

$$\mathbf{c}(i, m, n) = \max\left(\hat{\mathbf{c}}(i, m, n) - \frac{1}{\tau} \iota(i, n), 0\right) \quad (\text{norm\_flag} \neq 1) \quad (8)$$

or

$$\mathbf{c}(i, m, n) = \max\left(\hat{\mathbf{c}}(i, m, n) - \frac{1}{\tau} \iota(i, n), 0\right) \quad (\text{norm\_flag} = 1) \quad (9)$$

If no inhibition is specified then  $\iota(i, n) = 1$  ( $\forall i, n$ ).

The cross-correlograms  $\mathbf{C}$  for frame  $l$  are calculated by averaging only the inhibited cross-correlations within a given frame for which the corresponding IC value exceeds a threshold value  $\Theta_\chi$ :

$$\mathbf{C}(i, l, m) = \begin{cases} 0 & \text{if } \Psi = \emptyset \\ \frac{1}{|\Psi|} \sum_{d \in \Psi} \mathbf{c}(i, d, m) & \text{otherwise} \end{cases} \quad (10)$$

where  $\{\Psi \in n : (l - 1)M + 1 \leq n \leq lM, l \leq L - N + 1, \chi(i, n) \geq \Theta_\chi\}$  and  $\emptyset$  is the empty set. This can effectively be bypassed by setting  $\Theta_\chi = 0$ .

## Note

It is recommended that if `norm_flag = 1` then  $\tau \gg 1$ . As can be seen above, as  $\tau \rightarrow 1$  then  $\mathbf{a}_{\{L, R\}}(i, m, n - 1) \rightarrow 0$ , and hence  $\hat{\mathbf{c}}(i, m, n) \rightarrow 1$ .