Notes

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1 MASTER algorithm

Let $s_a \mathbf{I}^a(\hat{\mathbf{n}})$ be a spin- s_a field, where s_a can be 0 (i.e. 1 single component - e.g. T) or 2 (i.e. 2 components - e.g. (Q, U)). The observed map is:

$$_{s_a}\tilde{\mathbf{I}}^a(\hat{\mathbf{n}}) = w^a(\hat{\mathbf{n}}) \left[_{s_a} \mathbf{I}^a(\hat{\mathbf{n}}) + N^a(\hat{\mathbf{n}})\right],\tag{1}$$

where $w(\hat{\mathbf{n}})$ is the weights map. The harmonic coefficients of the observed map can be written as:

$$s_a \tilde{\mathbf{I}}_{\ell_1 m_1}^a = \sum_{\ell_2, m_2} s_a \mathsf{W}_{\ell_1 \ell_2, m_1 m_2}^a \cdot s_a \mathbf{I}_{\ell_2 m_2}^a, \tag{2}$$

where the mixing matrix is

$$s_a W_{\ell_1 \ell_2, m_1 m_2}^a \equiv (-1)^m \sum_{\ell_3, m_3} w_{\ell_3 m_3}^a \left[\frac{(2\ell_1 + 1)(2\ell_2 + 1)(2\ell_3 + 1)}{4\pi} \right]^{1/2} \begin{pmatrix} \ell_1 & \ell_2 & \ell_3 \\ -m_1 & m_2 & m_3 \end{pmatrix} s_a J_{\ell_1 \ell_2 \ell_3}$$
(3)

with

$${}_{0}\mathsf{J}_{\ell_{1}\ell_{2}\ell_{3}} = \left(\begin{array}{ccc} \ell_{1} & \ell_{2} & \ell_{3} \\ 0 & 0 & 0 \end{array}\right),\tag{4}$$

$${}_{2}\mathsf{J}_{\ell_{1}\ell_{2}\ell_{3}} = \begin{pmatrix} \ell_{1} & \ell_{2} & \ell_{3} \\ 2 & -2 & 0 \end{pmatrix} \frac{1}{2} \begin{pmatrix} 1 + (-1)^{\ell_{1}+\ell_{2}+\ell_{3}} & i \left[(-1)^{\ell_{1}+\ell_{2}+\ell_{3}} - 1 \right] \\ -i \left[(-1)^{\ell_{1}+\ell_{2}+\ell_{3}} - 1 \right] & 1 + (-1)^{\ell_{1}+\ell_{2}+\ell_{3}} \end{pmatrix}$$
 (5)

Let us define the pseudo-power-spectrum $\tilde{\mathsf{C}}_{\ell}^{ab}$

$$\tilde{\mathsf{C}}_{\ell}^{ab} \equiv \frac{1}{2\ell+1} \sum_{m} {}_{s_a} \mathbf{I}_{\ell m}^a \cdot \left({}_{s_b} \mathbf{I}_{\ell m}^b \right)^{\dagger} \tag{6}$$

The relation between the pseudo-power-spectrum and the true power spectrum C_ℓ^{ab} can be derived to be of the form

$$\langle \tilde{\mathbf{C}}_{\ell}^{ab} \rangle = \sum_{\ell'} \mathsf{M}_{\ell\ell'}^{s_a s_b} \cdot \mathbf{C}_{\ell'}^{ab}, \tag{7}$$

where the mode-coupling matrix $\mathsf{M}^{s_a s_b}_{\ell \ell'}$ takes the form:

• Case $s_a = s_b = 0$:

$$\langle \tilde{C}_{\ell}^{T_a T_b} \rangle = \sum_{\ell'} M_{\ell \ell'}^{00} C_{\ell'}^{T_a T_b} \tag{8}$$

with

$$M_{\ell\ell'}^{00} = \frac{2\ell' + 1}{4\pi} \sum_{\ell''} W_{\ell''}^{ab} \begin{pmatrix} \ell & \ell' & \ell'' \\ 0 & 0 & 0 \end{pmatrix}^2$$
 (9)

• Case $s_a = 0, s_b = 2$:

$$\left\langle \left(\begin{array}{c} \tilde{C}_{\ell}^{T_a E_b} \\ \tilde{C}_{\ell}^{T_a B_b} \end{array} \right) \right\rangle = \sum_{\ell'} \left(\begin{array}{c} M_{\ell\ell'}^{0+} & 0 \\ 0 & M_{\ell\ell'}^{0+} \end{array} \right) \cdot \left(\begin{array}{c} C_{\ell'}^{T_a E_b} \\ C_{\ell'}^{T_a B_b} \end{array} \right)$$
(10)

with

$$M_{\ell\ell'}^{0+} = \frac{2\ell'+1}{4\pi} \sum_{\ell''} W_{\ell''}^{ab} \begin{pmatrix} \ell & \ell' & \ell'' \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} \ell & \ell' & \ell'' \\ 2 & -2 & 0 \end{pmatrix}$$
 (11)

• Case $s_a = 2$, $s_b = 2$:

$$\left\langle \begin{pmatrix} \tilde{C}_{\ell}^{E_{a}E_{b}} \\ \tilde{C}_{\ell}^{E_{a}B_{b}} \\ \tilde{C}_{\ell}^{B_{a}B_{b}} \\ \tilde{C}_{\ell}^{B_{a}B_{b}} \end{pmatrix} \right\rangle = \sum_{\ell'} \begin{pmatrix} M_{\ell\ell'}^{++} & 0 & 0 & M_{\ell\ell'}^{--} \\ 0 & M_{\ell\ell'}^{++} & -M_{\ell\ell'}^{--} & 0 \\ 0 & -M_{\ell\ell'}^{--} & M_{\ell\ell'}^{++} & 0 \\ M_{\ell\ell'}^{--} & 0 & 0 & M_{\ell\ell'}^{++} \end{pmatrix} \cdot \begin{pmatrix} C_{\ell'}^{E_{a}E_{b}} \\ C_{\ell'}^{E_{a}B_{b}} \\ C_{\ell'}^{B_{a}B_{b}} \\ C_{\ell'}^{B_{a}B_{b}} \end{pmatrix} \tag{12}$$

with

$$M_{\ell\ell'}^{++} = \frac{2\ell' + 1}{4\pi} \sum_{\ell''} W_{\ell''}^{ab} \begin{pmatrix} \ell & \ell' & \ell'' \\ 2 & -2 & 0 \end{pmatrix}^2 \frac{1 + (-1)^{\ell + \ell' + \ell''}}{2}$$
(13)

$$M_{\ell\ell'}^{--} = \frac{2\ell' + 1}{4\pi} \sum_{\ell''} W_{\ell''}^{ab} \begin{pmatrix} \ell & \ell' & \ell'' \\ 2 & -2 & 0 \end{pmatrix}^2 \frac{1 - (-1)^{\ell + \ell' + \ell''}}{2}, \tag{14}$$

where in all these equations $W_{\ell''}^{ab}$ is the cross-spectrum of the weights map (without the $(2\ell+1)$ normalization):

$$W_{\ell}^{ab} \equiv \sum_{m} w_{\ell m}^{a} (w_{\ell m}^{b})^{*}. \tag{15}$$

Note that, in Eq. 7 one should add, on the right-hand side, the noise cross-power-spectrum:

$$\langle \tilde{\mathsf{N}}_{\ell}^{ab} \rangle \equiv \frac{1}{2\ell + 1} \sum_{m} \langle \mathbf{N}_{\ell m}^{a} \cdot (\mathbf{N}_{\ell m}^{b})^{\dagger} \rangle \tag{16}$$

1.1 Beam

Adding the effect of a beam amounts to redefining:

$$\mathsf{M}_{\ell_1\ell_2}^{s_as_b} \to \mathsf{M}_{\ell_1\ell_2}^{s_as_b} b_{\ell_2}^{ab},$$
 (17)

where b_{ℓ}^{ab} is the product of the harmonic transform of the beams for maps a and b.

1.2 Bandpowers

Consider the case where you want to compute the power spectrum in band-powers given by

$$\mathbf{B}_{k}^{ab} \equiv \frac{1}{N_{k}} \sum_{\ell=\ell_{k}}^{\ell_{k}+N_{k}-1} f(\ell) \mathbf{C}_{\ell}^{ab}, \tag{18}$$

then Eq. 7 above becomes

$$\langle \tilde{\mathbf{B}}_{k}^{ab} \rangle = \sum_{k'} \mathsf{M}_{kk'}^{B,s_{a}s_{b}} \cdot \mathbf{B}_{k'}^{ab} + \langle \tilde{\mathsf{N}}_{k}^{B,ab} \rangle \tag{19}$$

where the binned coupling matrix M^{B,s_as_b} is

$$\mathsf{M}_{k_1,k_2}^{B,s_as_b} \equiv \frac{1}{N_{k_1}} \sum_{\ell_1=\ell_{k_1}}^{\ell_{k_1}+N_{k_1}-1} \sum_{\ell_2=\ell_{k_2}}^{\ell_{k_2}+N_{k_2}-1} \frac{f(\ell_1)}{f(\ell_2)} \mathsf{M}_{\ell_1\ell_2}^{s_as_b} \tag{20}$$

2 Minimum variance quadratic estimator

Let \mathbf{d} be the data, given as a pixelized full-sky map, with covariance matrix

$$\langle \mathbf{d} \, \mathbf{d}^T \rangle \equiv C(\hat{\mu}) \equiv S(\hat{\mu}) + N(\hat{\mu}) = \sum_{\ell} C_{\ell} \, P_{\ell}(\hat{\mu}) + N(\hat{\mu}), \tag{21}$$

where $\hat{\mu} = \mathbf{n} \mathbf{n}^T$, \mathbf{n} is the vector of unit vectors containing the angular coordinates to each pixel and $P_{\ell}(x) \equiv (2\ell+1)L_{\ell}(x)/(4\pi)$, where L_{ℓ} are the Legendre polynomials. Note that we can expand P_{ℓ} as

$$P_{\ell}(\hat{\mu}) \equiv \sum_{m=-\ell}^{\ell} Y_{\ell m}(\mathbf{n}) \cdot Y_{\ell m}^{\dagger}(\mathbf{n}). \tag{22}$$

We will parametrize the power spectrum with a set of step functions, such that

$$C_{\ell} = \sum_{b} c_{b} \Theta_{b}(\ell), \quad \leftarrow \quad \Theta_{b}(\ell) = 1 \text{ if } \ell \in [\ell_{\min}^{b}, \ell_{\max}^{b}), \text{ 0 otherwise,}$$
 (23)

thus we can write

$$C(\hat{\mu}) = \sum_{b} c_b Q_b(\hat{\mu}) + N(\hat{\mu}), \text{ with } Q_b(\mu) = \sum_{\ell} \Theta_b(\ell) P_{\ell}(\mu).$$
 (24)

Let us write the most general quadratic estimator for the coefficients c_b :

$$\tilde{c}_b \equiv \mathbf{d}^T \hat{E}_b \mathbf{d} - B_b. \tag{25}$$

The mean value of \tilde{c}_b is:

$$\langle \tilde{c}_b \rangle = \sum_a c_a \text{Tr}(\hat{E}_b \, \hat{Q}_a) + \text{Tr}(\hat{E}_b \hat{N}) - B_b,$$
 (26)

and therefore we can remove the noise bias by defining

$$B_b \equiv \text{Tr}(\hat{E}_b \hat{N}). \tag{27}$$

The covariance matrix of this estimator is:

$$\langle (\tilde{c}_a - \langle \tilde{c}_a \rangle) (\tilde{c}_b - \langle \tilde{c}_b \rangle) \rangle = 2 \text{Tr}(\hat{C} \hat{E}_a \hat{C} \hat{E}_b). \tag{28}$$

Minimizing the variance of \tilde{c}_a under the constrain that the coupling matrix $W_{ab} \equiv \text{Tr}(\hat{E}_b \, \hat{Q}_a)$ have unit diagonal, we find the following minimization problem:

$$L = \text{Tr} \left[\hat{E}_b \hat{C} \hat{E}_b \hat{C} - 2(\lambda \hat{E}_b \hat{Q}_b - 1) \right], \tag{29}$$

where λ is the lagrange multiplier related to the constraint. The solution to this problem is:

$$\hat{E}_b = \frac{\hat{C}^{-1}\hat{Q}_b\hat{C}^{-1}}{\text{Tr}(\hat{C}^{-1}\hat{Q}_b\hat{C}^{-1}\hat{Q}_b)}.$$
(30)

Thus, a decoupled, unbiased and minimum-variance estimator, \hat{c}_b , for c_b can be found as:

$$\frac{\mathbf{d}^T \hat{C}^{-1} \hat{Q}_b \hat{C}^{-1} \mathbf{d} - \text{Tr}(\hat{C}^{-1} \hat{Q}_b \hat{C}^{-1} \hat{N})}{\text{Tr}(\hat{C}^{-1} \hat{Q}_b \hat{C}^{-1} \hat{Q}_b)} = \sum_a \hat{c}_a \frac{\text{Tr}(\hat{C}^{-1} \hat{Q}_b \hat{C}^{-1} \hat{Q}_a)}{\text{Tr}(\hat{C}^{-1} \hat{Q}_b \hat{C}^{-1} \hat{Q}_b)}.$$
 (31)

Note that this estimator requires a guess for the covariance matrix of the data $C(\hat{\mu})$, and exact minimum variance will only be attained for the true covariance (which itself depends on the power spectrum coefficients c_b). The choice of prior covariance will define the type of estimator. Note that this formalism immediately encompasses cut-skies by setting to zero all elements of the full covariance matrix involving unobserved pixels (this would correspond to the limit of infinite uncorrelated noise for those pixels).

2.1 Strategies for the different terms

• \hat{C}^{-1} d. This term can be computed by solving the linear system:

$$\hat{C}\mathbf{z} = \mathbf{d} \tag{32}$$

via conjugate gradients. The action of $\hat{C} = \hat{S} + \hat{N}$ can be computed as follows:

$$[\hat{S}\mathbf{z}](\hat{\mathbf{n}}_1) \equiv \sum_{\hat{\mathbf{n}}_2} S(\hat{\mathbf{n}}_1 \cdot \hat{\mathbf{n}}_2) z(\hat{\mathbf{n}}_2)$$
(33)

$$= \frac{1}{\Delta\Omega} \int d\hat{\mathbf{n}}_2 \sum_{\ell m} C_\ell Y_{\ell m}^*(\hat{\mathbf{n}}_1) Y_{\ell m}^*(\hat{\mathbf{n}}_1) \, \tilde{z}(\hat{\mathbf{n}}_2)$$
(34)

$$= \frac{1}{\Delta\Omega} \sum_{\ell m} Y_{\ell m}^*(\hat{\mathbf{n}}_1) C_{\ell} \tilde{z}_{\ell m} \tag{35}$$

$$= \frac{1}{\Delta\Omega} SHT^{-1} \left[C_{\ell} SHT[\tilde{z}(\hat{\mathbf{n}})]_{\ell m} \right], \tag{36}$$

where \tilde{z} is the extension of z to the whole sphere, with $\tilde{z} = 0$ in all unobserved pixels.

In most cases the noise power spectrum will be white, in which case $\hat{N}\mathbf{z}$ is just $\sum_{\hat{\mathbf{n}}} \sigma_N^2(\hat{\mathbf{n}}) z(\hat{\mathbf{n}})$, with $\sigma^2(\hat{\mathbf{n}})$ the per-pixel variance.

• $\mathbf{d}^T \hat{C}^{-1} \hat{Q}_b \hat{C}^{-1} \mathbf{d}$. With \mathbf{z} defined as above, this is just:

$$\mathbf{z}^T \hat{Q}_b \mathbf{z} = \frac{1}{(\Delta \Omega)^2} \sum_{\ell} \Theta_b(\ell) \sum_{m} |\tilde{z}_{\ell m}|^2$$
(37)