## Notes EESL

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October 26, 2020

## 1 Analytical Calculation

Let us first remind ourselves of the task at hand. Given a total recorded signal J(E) with a known zero-loss peak Z(E), we want to recover the single scattering distribution S(E) via deconvolution. Note that these are related by

$$J(E) = Z(E) + J^{1}(E) + J^{2}(E) + J^{3}(E) + \dots$$

$$= Z(E) + \frac{Z(E)}{I_{0}} * S(E) + \frac{Z(E)}{2!I_{0}^{2}} * S(E) * S(E) + \dots$$

$$\equiv Z(E) + J_{\text{in}}(E), \tag{1}$$

where we have defined  $J_{in}(E)$  to absorb all the inelastic scattering contributions. Upon transforming eq. (1) to Fourier space, the convolutions become products and we obtain

$$j(\nu) = z(\nu) + \frac{z(\nu)s(\nu)}{I_0} + \frac{z(\nu)s(\nu)^2}{2!I_0^2} + \dots$$

$$= z(\nu) \left[ 1 + \frac{s(\nu)}{I_0} + \frac{1}{2!} \left( \frac{s(\nu)}{I_0} \right)^2 + \dots \right]$$

$$= z(\nu) \exp \left[ \frac{s(\nu)}{I_0} \right]$$
(2)

From the definition of  $J_{\rm in}(E)$  in eq. (1) we also have

$$j(\nu) = z(\nu) + j_{\rm in}(\nu). \tag{3}$$

Combining eq. (2) and (3) therefore gives

$$j_{\rm in}(\nu) + z(\nu) = z(\nu) \exp\left[\frac{s(\nu)}{I_0}\right] \implies s(\nu) = I_0 \log\left[1 + \frac{j_{\rm in}(\nu)}{z(\nu)}\right].$$
 (4)

In our toy model we take Gaussians for both J(E) and Z(E), that is

$$J_{\rm in}(E) = \frac{I_s}{\sqrt{2\pi\sigma_s^2}} \exp\left[-\frac{(E - E_s)^2}{2\sigma_s^2}\right]$$
 (5)

$$Z(E) = \frac{I_0}{\sqrt{2\pi\sigma_B^2}} \exp\left[-\frac{E^2}{2\sigma_B^2}\right]. \tag{6}$$

Here, the zero-loss peak Z(E) and the inelastic scattering distribution  $J_{\rm in}$  are centered about zero and  $E_s$  respectively. In addition, we have  $\sigma_B \ll \sigma_s$ , meaning that the background has a much tighter peak than the signal by various orders of magnitude.

In the following we will need the Fourier transform of eq. (5) and (6). They are

$$j_{\rm in}(\nu) = I_s \exp[2\pi i \nu E_s] \exp[-2\pi^2 \sigma_s^2 \nu^2] \tag{7}$$

$$z(\nu) = I_0 \exp[-2\pi^2 \sigma_B^2 \nu^2]. \tag{8}$$

The idea is now to find S(E) by expanding eq. (4) and taking the inverse Fourier transform of each term in the expansion. Let us first show how to expand eq. (4). After defining  $g(\nu) \equiv j_{\rm in}(\nu)/z(\nu)$ , we get

$$s(\nu) = I_0 \left( g(\nu) - \frac{g(\nu)^2}{2!} + \frac{g(\nu)^3}{3!} - \dots \right).$$
 (9)

Note that each term in this expansion corresponds to a Gaussian, with subsequent terms getting an increasingly smaller spread. To be specific, from eq. (7) and (8) we find

$$g(\nu) = \frac{I_s}{I_0} \exp\left[-2\pi^2(\sigma_s^2 - \sigma_B^2)\nu^2 + 2\pi i\nu E_s\right]$$

$$= \frac{I_s}{I_0} \exp\left[-\frac{E_s^2}{2(\sigma_s^2 - \sigma_B^2)}\right] \exp\left[-2\pi^2(\sigma_s^2 - \sigma_B^2)\left(\nu - \frac{iE_s}{2\pi(\sigma_s^2 - \sigma_B^2)}\right)^2\right], \quad (10)$$

where we have completed the square on the last line. If we furthermore define  $\sigma_g^2 \equiv \sigma_s^2 - \sigma_B^2$ , the  $k^{\rm th}$  order contribution to  $s(\nu)$  from eq. (9) can be written as

$$g^{k}(\nu) = \left(\frac{I_s}{I_0}\right)^k \exp\left[-\frac{kE_s^2}{2\sigma_g^2}\right] \exp\left[-2k\pi^2\sigma_g^2\left(\nu - \frac{iE_s}{2\pi\sigma_g^2}\right)^2\right]. \tag{11}$$

Next, we take the inverse Fourier transform of  $g^k(\nu)$  and shift  $\nu \to \nu + iE_s/2\pi\sigma_g^2$ :

$$G_{k}(E) = \left(\frac{I_{s}}{I_{0}}\right)^{k} \exp\left[-\frac{kE_{s}^{2}}{2\sigma_{g}^{2}}\right] \mathcal{F}^{-1} \left\{ \exp\left[-2k\pi^{2}\sigma_{g}^{2}\left(\nu - \frac{iE_{s}}{2\pi\sigma_{g}^{2}}\right)^{2}\right] \right\}$$

$$= \left(\frac{I_{s}}{I_{0}}\right)^{k} \exp\left[-\frac{kE_{s}^{2}}{2\sigma_{g}^{2}}\right] \exp\left[\frac{E_{s}E}{\sigma_{g}^{2}}\right] \mathcal{F}^{-1} \left\{ \exp\left[-2k\pi^{2}\sigma_{g}^{2}\nu^{2}\right] \right\}$$

$$= \left(\frac{I_{s}}{I_{0}}\right)^{k} \exp\left[-\frac{kE_{s}^{2}}{2\sigma_{g}^{2}}\right] \exp\left[\frac{E_{s}E}{\sigma_{g}^{2}}\right] \frac{1}{\sqrt{2k\pi\sigma_{g}^{2}}} \exp\left[-\frac{E^{2}}{2k\sigma_{g}^{2}}\right]. \tag{12}$$

As can be seen from (12) we get another Gaussian for  $G_k(E)$ . This can be made explicit by completing the square:

$$G_k(E) = \frac{1}{\sqrt{2k\pi\sigma_g^2}} \left(\frac{I_s}{I_0}\right)^k \exp\left[-\frac{kE_s^2}{2\sigma_g^2}\right] \exp\left[-\frac{(E - kE_s)^2}{2k\sigma_g^2}\right] \exp\left[\frac{kE_s^2}{2\sigma_g^2}\right]$$

$$= \frac{1}{\sqrt{2k\pi\sigma_g^2}} \left(\frac{I_s}{I_0}\right)^k \exp\left[-\frac{(E - kE_s)^2}{2k\sigma_g^2}\right]. \tag{13}$$

Hence, the single scattering distribution S(E) becomes

$$S(E) = I_0 \left( G_1(E) - \frac{G_2(E)}{2!} + \frac{G_3(E)}{3!} - \dots \right)$$

$$= \frac{I_s}{\sqrt{2\pi(\sigma_s^2 - \sigma_B^2)}} \exp\left[ -\frac{(E - E_s)^2}{2(\sigma_s^2 - \sigma_B^2)} \right] - \frac{I_s^2/I_0}{2\sqrt{4\pi(\sigma_s^2 - \sigma_B^2)}} \exp\left[ -\frac{(E - 2E_s)^2}{4(\sigma_s^2 - \sigma_B^2)} \right] + \dots$$
(14)

As a sanity check, we notice that  $S(E) \to J_{\rm in}(E)$  if  $I_0 \gg I_s$  and  $\sigma_s \gg \sigma_B$ .