Supplement 1

Case 1: Centered Two-Sided Truncation Provided both x_a and x_b are symmetrically allocated around the mean of P and F_1 , it follows that i and k (Eqs. 4 and 5) are reduced to

$$i = 0;$$

$$k = \frac{\beta \phi(\beta) - \alpha \phi(\alpha)}{\Phi(\beta) - \Phi(\alpha)}.$$
(S1.1)

Correspondingly Eq. 10 is simplified to

$$\frac{pr}{P(Y)} = \Phi\left(\frac{\beta}{\sqrt{1 - 0.5Bkh^4}}\right) - \Phi\left(\frac{\alpha}{\sqrt{1 - 0.5Bkh^4}}\right).$$
(S1.2)

Due to the symmetry of the cdf, the above equation yields

$$\frac{pr}{P(Y)} = \left(1 - \Phi\left(\frac{\alpha}{\sqrt{1 - 0.5Bkh^4}}\right)\right) - \Phi\left(\frac{\alpha}{\sqrt{1 - 0.5Bkh^4}}\right),\tag{S1.3}$$

which can be further simplified as

$$\Phi^{-1}\left(0.5 - \frac{pr}{2P(Y)}\right) = \frac{\alpha}{\sqrt{1 - 0.5Bkh^4}},$$
(S1.4)

where $\Phi^{-1}(x)$ is the quantile function (inverse of the cdf) and h^2 is then calculated as

$$h^{2} = \frac{\sqrt{2}}{|Q|} \left(\frac{Q^{2} - \alpha^{2}}{Bk}\right)^{\frac{1}{2}},$$
(S1.5)

where $Q = \Phi^{-1} \left(0.5 - \frac{pr}{2P(Y)} \right)$.

Case 2: Right-Sided Truncation When $x_b \rightarrow +\infty$ (Fig. 1), *i* and *k* (Eqs. 4 and 5) reduce to

$$i = \frac{\phi(\alpha)}{1 - \Phi(\alpha)};$$

$$k = i(i - \alpha).$$
(S1.6)

Under these circumstances, i and k correspond to the standard tabulated values for the right-truncated selection (e.g. (Falconer and Mackay 1996)) and the rearrangement of Eq. 10 leads to

$$\frac{pr}{P(Y)} = 1 - \Phi\left(\frac{\alpha - Bih^2}{\sqrt{1 - 0.5Bkh^4}}\right),\tag{S1.7}$$

and

$$\Phi^{-1}\left(1 - \frac{pr}{P(Y)}\right) = \frac{\alpha - Bih^2}{\sqrt{1 - 0.5Bkh^4}}.$$
(S1.8)

Further arrangement leads to

$$h^{2} = \frac{2\alpha Bi \pm \left[4\alpha^{2}B^{2}i^{2} - 2\left(2B^{2}i^{2} + BQ'^{2}k\right)\left(\alpha^{2} - Q'^{2}\right)\right]^{\frac{1}{2}}}{(2B^{2}i^{2} + BQ'^{2}k)},$$
(S1.9)

where $Q' = \Phi^{-1}\left(1 - \frac{pr}{P(Y)}\right)$. There are two solutions of the above equation and verification using Eq. S1.7 provides the unique one.

Variance of the h^2 *estimate* Here we use the general h^2 estimate (Eq. 10) and further, we assume that p, i, and k are constants as they are all functions of the known input parameters x_a , x_b , μ , and σ_p^2 . Thus, to evaluate the variance of the h^2 estimate, we assume that h^2 is the exclusive function of r. Utilizing the maximum likelihood and linearization methods, we derived the following standard errors.

The standard error associated of the general h^2 estimate (Eq. 10) is

s.e.{
$$h^{2}$$
} = $\frac{\left(1 - 0.5h^{4}kB\right)^{\frac{3}{2}}}{\left|\phi\left(b\right)\left(-iB + 0.5h^{2}\beta kB\right) - \phi\left(a\right)\left(-iB + 0.5h^{2}\alpha kB\right)\right|} \frac{p}{P(Y)}\sqrt{r(1-r)}\frac{1}{\sqrt{n}},$ (S1.10)

where *n* is the number of progeny in $p, h^2 \neq 0$ under symmetric two-sided truncation, and *a* and *b* are

$$a = \frac{\alpha - Bih^2}{\sqrt{1 - 0.5Bkh^4}},$$

$$b = \frac{\beta - Bih^2}{\sqrt{1 - 0.5Bkh^4}}.$$
(S1.11)

The standard error of the h^2 estimate for the scenario of right-sided truncation (Eq. 12) is

s.e.{
$$h^{2}$$
} = $\frac{(1 - 0.5h^{4}kB)^{\frac{3}{2}}}{|-\phi(a)(-iB + 0.5h^{2}\alpha kB)|} \frac{p}{P(Y)}\sqrt{r(1-r)}\frac{1}{\sqrt{n}}.$ (S1.12)