

## Additional File 2 - Conjugate prior distributions

We use the Normal-Inverse- $\chi^2$  distribution as a prior on the parameters of model distribution of the bicluster-array combination  $(\mu_{a,b}, \sigma_{a,b})$ . This defines the following prior distributions for the mean and standard deviation:

$$\sigma \sim \text{Inv-}\chi^2(v_0, \sigma_0^2) \quad (2.1)$$

$$\mu | \sigma \sim N(\mu_0, \sigma^2 / \kappa_0) \quad (2.2)$$

This leads to the following formal distribution of the prior which is parameterized by

$(\mu_0, \kappa_0, v_0, \sigma_0^2)$  :

$$P(\mu_{a,b}, \sigma_{a,b}) = \sigma^{-1} (\sigma^2)^{-v_0/2+1} \exp\left(-\frac{1}{2\sigma^2} [v_0\sigma_0^2 + \kappa_0(\mu_0 - \mu)^2]\right) \quad (2.3)$$

The posterior distribution  $P(E.\text{level}, G.B.A.B.A.ID)$ .  $P(\mu_{a,b}, \sigma_{a,b})$  is then again a Normal-Inverse- $\chi^2$  distribution, characterized by:

$$\mu_p = \frac{\kappa_0}{\kappa_0 + n_X} \mu_0 + \frac{\kappa_0}{\kappa_0 + n_X} \frac{S_X}{n_X} \quad (2.4)$$

$$\kappa_p = \kappa_0 + n_X \quad (2.5)$$

$$v_p = v_0 + n_X \quad (2.6)$$

$$v_p \sigma_p^2 = v_0 \sigma_0^2 + (n_X - 1) ssq + \frac{\kappa_0 n_X}{\kappa_0 + n_X} \left(\frac{S_X}{n_X} - \mu_0\right)^2 \quad (2.7)$$

As shown in Equation (2.3), four hyperparameters  $(\mu_0, \kappa_0, v_0, \sigma_0^2)$  determine the prior distribution.