## 5.7 Gram-Charlier

A normalized form for the assumed distribution of the data x is

$$f(x,\sigma,\alpha) = \frac{1}{\sqrt{2\pi}\sigma} e^{\frac{-x^2}{2\sigma^2}} \left( 1 + \alpha \left(1 - \left(\frac{x}{2}\right)^2\right) \right)$$

where  $\sigma$  is the standard deviation of the basic Gaussian and  $\alpha$  measures the contribution of the Gram-Charlier term.

For a set of N data  $X_i$  the likelihood function is a product, so that

$$\operatorname{prob}(\alpha, \sigma | X_i) \propto \prod_i f(X_i, \sigma, \alpha).$$

Because f is a sum, the product gets very complicated. To investigate further, assume  $\alpha << 1$  and keep only terms linear in the product. This gives

$$\operatorname{prob}(\alpha, \sigma | X_i) \propto \frac{1}{(\sqrt{2\pi}\sigma)^N} \left( e^{\frac{-\sum_i X_i^2}{2\sigma^2}} + \alpha \sum_k (1 - X_k^2) e^{\frac{-\sum_{i \neq k} X_i^2}{2\sigma^2}} \right) \operatorname{prob}(\alpha) \operatorname{prob}(\sigma)$$

Marginalizing out  $\alpha$  can be more or less complicated to taste – the prior is open to debate. Taking a uniform distribution between 0 and t for  $\alpha$ , so prob $(\alpha) = 1/t$ , the relevant factors are simple, giving

$$\operatorname{prob}(\sigma|X_i) \propto \left(e^{\frac{-\sum_i X_i^2}{2\sigma^2}} + \frac{t}{2} \sum_k (1 - X_k^2) e^{\frac{-\sum_{i \neq k} X_i^2}{2\sigma^2}}\right) \operatorname{prob}(\sigma).$$

The Jeffreys prior is appropriate for  $\operatorname{prob}(\sigma)$ . Graphing the distribution for small sets of Gaussian data, taking  $t \simeq 0.1$ , we see a definite tendency for smaller values of the most likely  $\sigma$  than we get for  $\alpha = 0$ . The extra term in the Gram-Charlier expansion has a tendency to absorb some of the spread in the data.

The odds on including the Gram-Charlier term are given by the ratio

$$\mathcal{O} = \frac{\int \operatorname{prob}(\sigma | X_i, \operatorname{finite} \alpha) d\sigma}{\int \operatorname{prob}(\sigma | X_i, \alpha = 0) d\sigma}$$

which is the ratio of the weights of evidence. These integrations can be done numerically. For smallish sets of data  $(N \simeq 30)$  we find typical odds of about 10 to 1 (assuming equal prior probabilities for the two hypotheses) in favour of including the term. While the effect on the distribution of small values of  $\alpha$  may seem limited, it can affect significances in the tails by a per cent. or so.