## APPENDIX 1

Following Schreiber et al (2003), we created the reliability statistic as follows.

Let  $\{x_i\}$  and  $\{y_i\}$  be the spike times from two trials with n and m spikes respectively. We can turn the  $\{x_i\}$  into a continuous function by allowing each spike time to be a Dirac delta function about  $x_i$ . Thus

$$(1) \{x_i\} \rightarrow \sum_{i=1}^n \delta (t - x_i)$$

By convolving with a normalized Gaussian of width  $\sigma$ , we turn the discrete spike times into a well behaved, smooth curve.

$$x (t) = \int_{-\infty}^{\infty} \sum_{i=1}^{n} \delta (\tau - x_{i}) \frac{1}{\sqrt{2 \pi} \sigma} e^{\frac{-(t-\tau)^{2}}{2 \sigma^{2}}} d\tau$$

$$(2) \qquad x (t) = \frac{1}{\sqrt{2 \pi} \sigma} \sum_{i=1}^{n} e^{\frac{-(t-x_{i})^{2}}{2 \sigma^{2}}}$$

Similarly

(3) 
$$y(t) = \frac{1}{\sqrt{2\pi}\sigma} \sum_{j=1}^{m} e^{\frac{-(t-y_j)^2}{2\sigma^2}}$$

Schreiber et al (2003) defines the reliability between two spike trains as the normalized dot product of these functions.

$$(4) R_{xy} = \frac{x \cdot y}{|x| |y|}$$

However, whereas Schreiber et al (2003) advocates treating x and y as vectors, and calculating the dot product point by point over an arbitrary  $\Delta t$ , we chose to keep them as continuous functions.

(5) 
$$x \cdot y = \int_{-\infty}^{\infty} x (t) y (t) dt$$

$$= \int_{-\infty}^{\infty} \left( \sum_{i=1}^{n} \frac{1}{\sqrt{2 \pi} \sigma} e^{\frac{-(t-x_{i})^{2}}{2 \sigma^{2}}} \right) \left( \sum_{j=1}^{m} \frac{1}{\sqrt{2 \pi} \sigma} e^{\frac{-(t-y_{j})^{2}}{2 \sigma^{2}}} \right) dt$$

$$(6) \qquad x \cdot y = \frac{1}{2 \pi \sigma^{2}} \sum_{j=1}^{n} \int_{-\infty}^{\infty} e^{\frac{-(t-x_{i})^{2} - (t-y_{j})^{2}}{2 \sigma^{2}}} dt$$

Manipulating the numerator of the exponent in (6) by expanding the terms and completing the square yields:

$$x \cdot y = \frac{1}{2 \pi \sigma^{2}} \sum_{i=1}^{n} \sum_{j=1}^{m} \int_{-\infty}^{\infty} e^{\frac{-(x_{i} - y_{j})^{2}}{4 \sigma^{2}}} e^{\frac{-(t - \frac{x_{i} + y_{j}}{2})^{2}}{\sigma^{2}}} dt$$

$$(7) \qquad x \cdot y = \frac{1}{2 \sqrt{\pi} \sigma} \sum_{i=1}^{n} \sum_{j=1}^{m} e^{\frac{-(x_{i} - y_{j})^{2}}{4 \sigma^{2}}}$$

We can now write down the norms by defining  $|x| = \sqrt{x \cdot x}$  and  $|y| = \sqrt{y \cdot y}$ .

(8) 
$$|x| = \sqrt{x \cdot x} = \sqrt{\left(\sum_{j=1}^{n} \sum_{j=1}^{n} e^{\frac{-(x_{i} - x_{j})^{2}}{4\sigma^{2}}}\right)}$$

$$| y | = \sqrt{y \cdot y} = \sqrt{ \left[ \sum_{i=1}^{m} \sum_{j=1}^{m} e^{\frac{-(y_i - y_j)^2}{4\sigma^2}} \right] }$$

Finally, we can insert (7-9) into (4) and explicitly write the reliability in terms of spike times and  $\sigma$ .

(10) 
$$R_{xy} = \frac{\sum_{j=1}^{n} \sum_{j=1}^{m} e^{\frac{-(x_{i} - y_{j})^{2}}{4 \sigma^{2}}}}{\sqrt{\left(\sum_{j=1}^{n} \sum_{j=1}^{n} e^{\frac{-(x_{i} - x_{j})^{2}}{4 \sigma^{2}}}\right) \left(\sum_{j=1}^{m} \sum_{j=1}^{m} e^{\frac{-(y_{i} - y_{j})^{2}}{4 \sigma^{2}}}\right)}$$

Although the expression appears bulky, it significantly speeds computation time when compared with treating x and y as vectors. Reliability across N trials can be found by averaging the pairwise reliabilities:

(11) 
$$R = \frac{2}{N(N-1)} \sum_{i=1}^{N-1} \sum_{j=i+1}^{N} R_{ij}$$