Introduction to the Transverse Spatial Correlations in Spontaneous Parametric Down-Conversion Through the Biphoton Birth Zone James Schneeloch^{1,2,3} and John C. Howell ^{1,2}



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 $\chi^{(2)}$ -Nonlinear Crystal Pump Field Fig. 1: Depiction of second-order nonlinear crystal axis. The direction of propagation here is the z-axis, while the transverse components are in the x and y directions. The z-axis is also called the optic axis. Fig. 2: Phase-Matching relationship for nearly collinear downconverted light. When θ is small, we may use the small-angle approximation to get a simple expression for our biphoton wavefunction. Pump Filte Fig. 3: Diagram of an experiment used to measure the transverse position correlations between signal and idler photons in type-I SPDC. A pair of lenses is used to image the face of the nonlinear crystal onto DMD (Digital Micro-mirror Device) arrays. By selecting pixels of these DMD arrays to reflect toward photo detectors, one obtains the joint position probabilities from the histogram of coincidence counts. **Fourier Transform Properties of** the Double-Gaussian The Heisenberg relations: $\sigma_{\chi_+}\sigma_{k_+} \geq \frac{1}{2}$ 1.0 -1.0 -0.5 0.5 $\sigma_{x_{-}}\sigma_{k_{-}} \geq \frac{-}{2}$ Are saturated by the Double-Gaussian state. In addition, the conditional Heisenberg relations: $\sigma_{x_1}\sigma_{(k_1|k_2)} \geq \frac{1}{2}$ TABLE I: Statistics of the Double Gaussian $\sigma_{(x_1|x_2)}\sigma_{k_1} \ge \frac{1}{2}$ Value $\langle x_1 \rangle = \langle x_2 \rangle = 0$ marginal means The **Birth Zone Number** $\,N$.. are also saturated by the Double-Gaussian state. $|\langle x_2 \rangle_{\rho(x_2|x_1)} = x_1 - \frac{2x_1 \sigma_x^2}{\sigma_{x_+}^2 + \sigma_{x_-}^2} = rx_1$ conditioned mean $=\left(\frac{\sigma_{(x_1+x_2)}}{\sigma_{(x_1+x_2)}}\right)$ $\sigma_{x_1}^2 = \sigma_{x_2}^2 = \frac{\sigma_{x_+}^2 + \sigma_{x_-}^2}{2}$ marginal variance Lower bound for birth zone size conditioned variance $\sigma_{(x_1|x_2)}^2 = \sigma_{(x_2|x_1)}^2 = \frac{2\sigma_{x_+}^2 \sigma_{x_-}^2}{\sigma_{x_+}^2 + \sigma_{x_-}^2}$ $\langle \sigma_{(x_1-x_2)} \rangle$ $\langle x_1 x_2 \rangle - \langle x_1 \rangle \langle x_2 \rangle = \frac{\sigma_{x_+}^2 - \sigma_{x_-}^2}{2}$ For: co-variance Since $\sigma_{(x_1-x_2)} \ge \sigma_{(x_1|x_2)}$, $r = \frac{\langle x_1 x_2 \rangle - \langle x_1 \rangle \langle x_2 \rangle}{\sigma_{x_1} \sigma_{x_2}} = \frac{\sigma_{x_+}^2 - \sigma_{x_-}^2}{\sigma_{x_+}^2 + \sigma_{x_-}^2}$ Pearson r value $h(x_1, x_2) = \log(2\pi e \sigma_{x_\perp} \sigma_{x_\perp})$ We know that joint entropy $h(x_1) = \frac{1}{2} \log(\pi e(\sigma_{x_+}^2 + \sigma_{x_-}^2))$ marginal entropy mutual information $h(x_1:x_2) = \log\left(\frac{\sigma_{x_+}^2 + \sigma_{x_-}^2}{2\sigma_{x_+}\sigma_{x_-}}\right)$ $\sigma_{(x_1 - x_2)} \ge \frac{1}{2\sigma_{k_1}}$ $| = \log \left(\frac{\sigma_{x_1}}{\frac{\sigma_{(x_1|x_2)}}{\sigma_{(x_1|x_2)}}} \right)$

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