Supplemental Material

Cosmic Bell Test: Measurement Settings from Milky Way Stars

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Causal Alignment

Compared to a standard Bell test, the time-dependent locations of the stars on the sky relative to our ground-based experimental sites complicate the enforcement of the space-like separation conditions needed to address both the locality and freedom-of-choice loopholes. For example, the photon from star $A_k$ must be received by Alice’s stellar photon receiving telescope (Rx-SP) before that photon’s causal wave front reaches either the Rx-SP or the entangled photon receiving telescope (Rx-EP) on Bob’s side, and vice versa.

To compute the time-dependent durations $\tau_{\text{valid}}^k(t)$ (for $k = (A, B)$) that settings chosen by astronomical photons remain valid, we adopt a coordinate system with the center of the Earth as the origin. The validity times on each side due to the geometric configuration of the stars and ground-based sites are then given by

$$\tau_{\text{valid}}^A(t) = \frac{1}{c} \hat{n}_S(t) \cdot (\vec{r}_A - \vec{m}_A) + \frac{\eta_A}{c} |\vec{m}_A - \vec{s} - |\vec{m}_B - \vec{s}||$$

$$\tau_{\text{valid}}^B(t) = \frac{1}{c} \hat{n}_S(t) \cdot (\vec{r}_B - \vec{m}_B) + \frac{\eta_B}{c} |\vec{m}_B - \vec{s} - |\vec{m}_A - \vec{s}||$$

where $\vec{r}_A$ is the spatial 3-vector for the location of Alice’s Rx-SP, $\vec{m}_A$ is the spatial 3-vector for Alice’s Rx-EP (and likewise for $\vec{r}_B$ and $\vec{m}_B$ on Bob’s side), $\vec{s}$ is the location of the entangled photon source, and $c$ is the speed of light in vacuum. The time-dependent unit vectors $\hat{n}_S(t)$, $\hat{n}_S(t)$ point toward the relevant stars, and are computed using astronomical ephemeris calculations. Additionally, $\eta$ is the index of refraction of air and $\eta_A, \eta_B$ parametrize the group velocity delay through fiber optics / electrical cables connecting the telescope and entangled photon detectors. To compute $\tau_{\text{valid}}^k(t)$, we make the reasonable approximation that the Rx-SP and Rx-EP are at the same spatial location on each side, such that $\vec{r}_A = \vec{m}_A$ and $\vec{r}_B = \vec{m}_B$, and the computations require the GPS coordinates of only 3 input sites (see Table I). This assumes negligible delays from fiber and electrical cables via the $\eta_A, \eta_B$ terms. Negative validity times $\tau_{\text{valid}}^k(t)$ for either side would indicate an instantaneous configuration that was out of “causal alignment,” in which at least one setting would be invalid for the purposes of closing the locality loophole. For runs 1 and 2, $\tau_{\text{valid}}^k(t) > 0$ for the entire duration of 179 s, with minimum times in Table I.

We subtract the time it takes to implement a setting with the electro-optical modulator, $\tau_{\text{set}} \approx 170$ ns, and subtract additional conservative buffer margins $\tau_{\text{buffer}}^k (0.38 \mu s$ for Alice and $1.76 \mu s$ for Bob) to determine the minimum time windows $\tau_{\text{used}}^k$ in Eq. (2) utilized during the experiment (see Table II):

$$\tau_{\text{used}}^k = \min_t \{\tau_{\text{valid}}^k(t) - \tau_{\text{buffer}} - \tau_{\text{set}}\}$$

where $\tau_{\text{set}}$ includes the total delays on either side due to reflections inside the telescope optics, the SPAD detector response, and electronic readout on the astronomical receiver telescope side as well as the time to switch the Pockels cell and electronically use the FPGA board to output a random number. The next section conservatively estimates $\tau_{\text{atm}} \approx 18$ ns for the delay due to the index of refraction of the atmosphere for either observer. While $\tau_{\text{atm}}$ is not explicitly considered in Eq. (2), it is well within the buffer margins, since $\tau_{\text{buffer}} \gg \tau_{\text{atm}}$, which also encompass any small inaccuracies in the timing or distances between the experimental sites.

Although $\tau_{\text{valid}}^k(t)$ depends on time, motivating our use of $\tau_{\text{valid}}^k \equiv \min_t \{\tau_{\text{valid}}^k(t)\}$ when computing $\tau_{\text{used}}^k$, the actual values of $\tau_{\text{valid}}^k$ changed very little during our observing windows. For the stars used in experimental run 1, $\Delta \tau_{\text{valid}}^A = 2.96$ ns and $\Delta \tau_{\text{valid}}^B = 17.26$ ns; for experimen-
Atmospheric Delay

The air in the atmosphere causes a relative delay between the causal light cone, which expands outward at speed $c$, and the photon, which travels at $c/n$, where $n$ is the index of refraction of air. We estimate this effect by computing the light’s travel time through the atmosphere on the way to the observer. If the atmosphere has index of refraction $n$ and scale height $z_0$, the delay time is

$$\Delta t = \frac{(z_0 - h)X(n - 1)}{c}$$

where $X$ is the airmass. The minimum elevation of each stellar photon receiving telescope is $h = 200$ m above sea level, and the minimum altitude angle is $\delta = 24^\circ$ above the horizon with airmass $X \approx 2.5$ (see Tables [I][II]). Neglecting the earth’s curvature (which is a conservative approximation), we use $z_0 = 8.0$ km and the index of refraction at sea level of $n - 1 \approx 2.7 \times 10^{-4}$ [I]. The delay between the arrivals of the causal light cone and the photon itself may be conservatively estimated to be $\Delta t = 17.6$ ns due to the atmosphere.

Source Selection

We used custom Python software to select candidate stars from the Hipparcos catalogue [2][3] with parallax distances greater than 500 ly, distance errors less than 50%, and Hipparcos $H_P$ magnitude between 5 and 9 to avoid detector saturation and ensure sufficient detection rates. Telescopes pointed out of open windows at both sites (see Table [I]). A list of ~100-200 candidate stars were pre-selected per side for each night due to the highly restrictive azimuth/altitude limits. Candidate stars were visible through the open windows for ~ 20-50 minutes on each side.

Due to weather, seeing conditions, and the uncertainties in aligning the transmitting and receiving telescope optics for the entangled photon source, it was not possible to pre-select specific star pairs for each experimental run at a predetermined time. Instead, when conditions were stable, we selected the best star pairs from our pre-computed candidate lists that were currently visible through both open windows, ranking stars based on brightness, distance, the amount of time each would remain visible, the settings validity time, and the airmass at the time of observation. The 4 bright stars we actually observed for runs 1 and 2 were ~5–6 mag (see Table [I]). Combined with the geometric configuration of the sites (see Table [I]), selection of these stars ensured sufficient setting validity times on both sides during each experimental run of 179 seconds.

<table>
<thead>
<tr>
<th>Site</th>
<th>Lat.°</th>
<th>Lon.°</th>
<th>Elev. [m]</th>
<th>Telescope [m]</th>
</tr>
</thead>
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<tr>
<td>Telescope A</td>
<td>48.21645</td>
<td>16.35431</td>
<td>215.0</td>
<td>0.2032</td>
</tr>
<tr>
<td>source $S$</td>
<td>48.22131</td>
<td>16.35643</td>
<td>205.0</td>
<td>...</td>
</tr>
<tr>
<td>Telescope B</td>
<td>48.23160</td>
<td>16.357953</td>
<td>200.0</td>
<td>0.254</td>
</tr>
</tbody>
</table>

TABLE I. Latitude, Longitude, Elevation, for Alice ($A$), Bob ($B$) and the Source ($S$), and aperture diameter of the stellar photon receiving telescopes.

Lookback Times

For stars within our own galaxy, the lookback time $t_k$ to a stellar emission event from a star $d_k$ light years away is $t_k = d_k$ years. For example, Hipparcos Star HIP 2876 is located $d_B = 3624$ light years (ly) from Earth, and its photons were therefore emitted $t_B = 3624$ years prior to us observing them (see Table [I]). The lookback time $t_E$ to when the past light cone of a stellar emission event from star $k$ intersects Earth’s worldline is $t_E = 2d_k$ years.

The lookback time to the past light cone intersection event $t_{AB}$ (in years) for a pair of Hipparcos stars is [4]

$$t_{AB} = \frac{1}{2} \left[ d_A + d_B + \sqrt{d_A^2 + d_B^2 - 2d_Ad_B\cos(\alpha)} \right],$$

where $d_A, d_B$ are the distances to the stars (in ly) and $\alpha$ is the angular separation (in radians) of the stars, as seen from Earth. See the lower left panel of Fig. [I].

Ignoring any covariance between $d_A, d_B$, and $\alpha$, and assuming the error on $\alpha (\sigma_{\alpha})$ is negligible compared to the distance errors ($\sigma_{d}$), the 1σ lookback time error is approximately given by

$$\sigma_{t_{AB}} \approx \sqrt{\sum_{i,j} \sigma_{d_i}^2 \left[ t_{AB} - \frac{d_i - d_j}{2} \right]^2},$$

where $(i, j) \in \{(A, B), (B, A)\}$.

Experimental Details

The entangled photon source was based on type–II spontaneous parametric down conversion (SPDC) in a periodically poled KTiOPO$_4$ (ppKTP) crystal with 25 mm length. Using a laser at 405 nm, the ppKTP crystal was bi-directionally pumped inside a polarization Sagnac interferometer generating degenerate polarization entangled photon pairs at 810 nm. We checked the performance of the SPDC source via local measurements at the beginning of each observation night. Singles and coincidence rates of approximately 1.1 MHz and 275 kHz, respectively, correspond to a local coupling efficiency (i.e., coincidence rate divided by singles rate)
of roughly 25%. In run 1 (run 2), the duty cycle of Alice’s and Bob’s measurements – i.e., the temporal sum of used valid setting intervals divided by the total measurement time per run – were 24.9% (22.0%) and 40.6% (44.6%), respectively, resulting in a duty-cycle for valid coincidence detections between Alice and Bob of 10.1% (9.8%). From the measured 136 332 (88 779) total valid coincidence detections per run, we can thus infer the total two-photon attenuation through the quantum channels to Alice and Bob of 15.3 dB (16.8 dB).

Quality of Setting Reader

The value of the observed CHSH violation is highly sensitive to the fraction of generated settings which were in principle “predictable” by a local hidden-variable model. For this reason, it is important to have a high-fidelity spectral model of the setting generation process. In our analysis, we conservatively assume that local noise and incorrectly generated settings are completely predictable and exploitable. An incorrectly generated setting is a red photon that generates a blue setting (or vice versa) by ending up at the wrong SPAD.

In this section we compute the fractions of incorrectly generated settings $f_{r\to b}$ and $f_{b\to r}$. For example, $f_{r\to b}$ is the conditional probability that a red photon goes the wrong way in the dichroic and ends up detected as a blue photon, generating the wrong setting. These fractions are highly sensitive to the transmission and reflection spectra of the two dichroic mirrors in each setting generator. They are somewhat less dependent on the spectral distribution of photons emitted by the astronomical source, on absorption and scattering in the Earth’s atmosphere, the anti-reflection coatings on the optics, and the SPAD quantum efficiencies.
A system of dichroic beamsplitters which generates measurement settings from photon wavelengths can be modeled by two functions \( \rho_{\text{red}}(\lambda) \) and \( \rho_{\text{blue}}(\lambda) \), the probability of transmission to the red and blue arms as a function of photon wavelength \( \lambda \). Ideally, photons with wavelength \( \lambda \) longer than some cutoff \( \lambda' \) would not arrive at the blue arm: \( \rho_{\text{blue}}(\lambda) = 0 \) for \( \lambda > \lambda' \). Similarly, \( \rho_{\text{red}}(\lambda) = 0 \) for \( \lambda \leq \lambda' \) would ensure that blue photons do not arrive at the red arm. Due to imperfect dichroic beamsplitters, however, it is impossible to achieve \( \rho_{\text{blue}}(\lambda > \lambda') = 0 \) and \( \rho_{\text{red}}(\lambda \leq \lambda') = 0 \).

The total number of blue settings generated by errant red photons can be computed as

\[
N_{\text{red-b}}(\lambda') = \int_{\lambda}^{\infty} \rho_{\text{blue}}(\lambda) N_{\text{in}}(\lambda) \, d\lambda,
\]

where \( N_{\text{in}}(\lambda) \) is the spectral distribution of the stellar photons remaining after losses due to the atmosphere, anti-reflection coatings, and detector quantum efficiency.

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\]

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Then the fraction \( f_{r\rightarrow b} \) can be computed by normalizing
\[
f_{r\rightarrow b} = \frac{N_{r\rightarrow b}(\lambda')}{N_{r\rightarrow b}(\lambda') + N_{r\rightarrow r}(\lambda')}.
\] (7)

We may then compute the \( \rho \)'s from measured dichroic mirror transmission and reflection curves and model \( N_{in}(\lambda) \). Finally, it is important to note that our red-blue color scheme is parametrized by the arbitrary cut-off wavelength \( \lambda' \). We may choose \( \lambda' \) to minimize the overall fraction of wrong settings,
\[
\lambda' = \arg \min \left\{ \frac{N_{r\rightarrow b} + N_{b\rightarrow r}}{N_{r\rightarrow r} + N_{r\rightarrow b} + N_{b\rightarrow r} + N_{b\rightarrow b}} \right\}.
\] (8)

For the four stars in the two observing runs, and the model of \( N_{in}(\lambda) \) described in the next section, the wrong-way fractions are tabulated in Table III. One typical analysis is illustrated in Fig. 4.

**Characterizing Dichroics**

Our setting reader uses a system of one shortpass (s) (Thorlabs M 254H45) and one longpass (l) (Thorlabs M 254C45) beamsplitter with transmission (T) and reflection (R) probabilities plotted in Fig. 4C. We choose to place the longpass beamsplitter in the reflected arm of the shortpass beamsplitter, instead of the other way around, to minimize the overall wrong-way fraction as written in Eq. (8). With this arrangement, \( \rho_{\text{blue}}(\lambda) = \rho_{T,s}(\lambda) \sim 10^{-3} \) for red wavelengths and \( \rho_{\text{red}}(\lambda) = \rho_{R,s}(\lambda)\rho_{T,l}(\lambda) \sim 10^{-3} \) for blue wavelengths. The transmission/reflection spectra of both dichroic mirrors and of the blue/red arms are plotted in Fig. 4C.

**Modeling the number distribution of photons**

In this section, we describe our model of \( N_{in}(\lambda) \), which covers the wavelength range 350 nm-1150 nm. We start with the stellar spectra, which can be modeled as blackbodies with characteristic temperatures taken from the Hipparcos catalogue [2, 3]. We then apply corrections for the atmospheric transmission \( \rho_{\text{atm}}(\lambda) \), two layers of anti-reflection coatings in each arm \( \rho_{\text{temp}}(\lambda) \), a silvered mirror \( \rho_{\text{mirror}} \), and finally the detector’s quantum efficiency \( \rho_{\text{det}}(\lambda) \) as the photon makes its way through the setting reader.

**Stellar Spectra**

As discussed in the main text, the stars were selected on the basis of their brightness, with temperatures ranging from 3150 K-7600 K. To a very good approximation, the photons emitted by the stars follow a blackbody distribution, which we assume is largely unaltered by the interstellar medium as the light travels towards earth:
\[
N_{\text{star}}(\lambda) = \frac{2c}{\lambda^3} \frac{1}{\exp(hc/(k\lambda T)) - 1}.
\] (9)

**Atmospheric Absorbance**

We generate an atmospheric transmittance spectrum with the MODTRAN model for mid-latitude atmospheres looking towards zenith [5]. To correct for the observation airmass (up to \( X = 2.5 \)), we use optical densities from [6] to compute the atmospheric transmission efficiency, which is due mostly to broadband Rayleigh scattering. A more sophisticated model could also compute modified absorption lines at higher airmasses, but the effect on the wrong-way fractions \( f_{r\rightarrow b}, f_{b\rightarrow r} \) is negligible compared to the spectral change resulting from Rayleigh scattering.

**Lenses and Detectors**

In the experimental setup, one achromatic lens in each arm collimates the incident beam of stellar photons. The collimated beam reflects off a silver mirror and is focused by a second lens onto the active area of the SPADs. These elements are appropriately coated in the range from 500 nm-1500 nm for minimum losses. However, not all photons are transmitted through the two lenses and the mirror. Each component has a wavelength-dependent probability of transmission that is close to unity for most of the nominal range, as plotted in Fig. 5. Once the focused light is incident on the SPAD, it will actually detect the photon with some wavelength-dependent quantum efficiency. The cumulative effect of these components on the incident spectrum is shown in Fig. 4B.

<table>
<thead>
<tr>
<th>Run</th>
<th>Side</th>
<th>HIP ID</th>
<th>( f_{r\rightarrow b} )</th>
<th>( f_{b\rightarrow r} )</th>
<th>Efficiency</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>36127</td>
<td>0.0142</td>
<td>0.0192</td>
<td>25.0%</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>105229A</td>
<td>0.0180</td>
<td>0.0146</td>
<td>24.9%</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>80620</td>
<td>0.0139</td>
<td>0.0203</td>
<td>24.3%</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>2876</td>
<td>0.0139</td>
<td>0.0160</td>
<td>22.7%</td>
</tr>
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</table>
FIG. 4. (A) Blackbody spectra of the stars used in Run 1, extincted by atmospheric Rayleigh scattering and telluric absorption, plotted in number flux per wavelength. (B) Our maximally conservative model of the anti-reflection coatings in the two lenses, the silver mirror, and detector quantum efficiency curves as a function of photon wavelength $\lambda$. (C) Which-way probabilities as a function of $\lambda$ due to the dichroic beamsplitters. Note that the addition of the longpass beamsplitter makes $\rho_{\text{red}}(\lambda)$ exceptionally flat, i.e. very good at rejecting blue photons. (D) Color distribution of photons seen at each arm are plotted, i.e. $N_{\text{in}}(\rho_{\text{red}})\text{ and } N_{\text{in}}(\rho_{\text{blue}})$. The curves are normalized so that the total area under the sum of both curves is 1. The color scheme’s cutoff wavelength $\lambda'$ is depicted by the shading color, and for this star is about $\lambda' \sim 703.2$ nm. Note that some of the photons arriving at each arm are classified as the wrong color (overlap of red and blue arm spectra), no matter which $\lambda'$ is chosen.

Data Analysis

In this section we analyze the data from the two experimental runs. We make the assumptions of fair sampling and fair coincidences [7]. Thus, for testing local realism, all data can be postselected to coincidence events between Alice’s and Bob’s measurement stations. These coincidences were identified using a time window of 2.5 ns.

We denote by $N_{ij}^{AB}$ the number of coincidences in which Alice had outcome $A \in \{+,-\}$ under setting $a_i$ ($i = 1, 2$) and Bob had outcome $B \in \{+,-\}$ under setting $b_j$ ($j = 1, 2$). The measured coincidences for run 1 were

<table>
<thead>
<tr>
<th>$i j \backslash AB$</th>
<th>++</th>
<th>+−</th>
<th>−+</th>
<th>−−</th>
</tr>
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<tbody>
<tr>
<td>11</td>
<td>2495</td>
<td>6406</td>
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<td>12</td>
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<td>22</td>
<td>18451</td>
<td>3512</td>
<td>3949</td>
<td>14196</td>
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</table>

The coincidence numbers for run 2 were

<table>
<thead>
<tr>
<th>$i j \backslash AB$</th>
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<th>+−</th>
<th>−+</th>
<th>−−</th>
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<tbody>
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<td>22</td>
<td>5359</td>
<td>1012</td>
<td>1249</td>
<td>4495</td>
</tr>
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</table>

We can define the number of all coincidences for setting combination $a_i b_j$,

$$N_{ij} = \sum_{A,B=+,−} N_{ij}^{AB},$$

and the total number of all recorded coincidences,

$$N = \sum_{i,j=1,2} N_{ij}.$$

A point estimate gives the joint setting choice probabilities

$$q_{ij} = p(a_i b_j) = \frac{N_{ij}}{N}.$$  

We first test whether the probabilities $q_{ij}$ can be factorized, i.e., that they can be (approximately) written as

$$p_{ij} = p(a_i) p(b_j).$$

Otherwise, there could be a common cause and the setting choices would not be independent. We define
\[ p(a_1) = 0.6193, \quad p(a_2) = 0.3807, \quad p(b_1) = 0.2257, \quad p(b_2) = 0.7743. \]

Pearson’s \( \chi^2 \)-test for independence, \( q_{ij} = p_{ij} \), yields \( \chi^2 = N \sum_{i,j=1}^2 (q_{ij} - p_{ij})^2 / p_{ij} = 1.132 \). This implies that, under the assumption of independent setting choices, there is a purely statistical chance of 0.287 that the observed data \( q_{ij} \) (or data even more deviating) are obtained. This probability is much larger than any typically used threshold for statistical significance. Hence, this test does not allow a refutation of the assumption of independent setting choices. For run 2, we estimate \( p(a_1) = 0.7333 \), \( p(a_2) = 0.2667 \), \( p(b_1) = 0.4854 \), and \( p(b_2) = 0.5146 \), with \( \chi^2 = 1.158 \) and statistical chance 0.282.

We next estimate the conditional probabilities for correlated outcomes in which both parties observe the same result:

\[ p(A = B|a_1,b_1) = \frac{N_{++} + N_{--}}{N_{ij}}, \]

The Clauser-Horne-Shimony-Holt (CHSH) inequality can be written as

\[ C = -p(A = B|a_1,b_1) - p(A = B|a_1,b_2) - p(A = B|a_2,b_1) + p(A = B|a_2,b_2) \leq 0. \]  

While the local-realist bound is 0, the quantum bound is \( \sqrt{2} - 1 = 0.414 \), and the logical (algebraic) bound is 1.

With our data, the CHSH values are \( C = 0.2125 \) for run 1, and \( C = 0.2509 \) for run 2, in each case violating the local-realist bound of zero. See Fig. 6. The widely known CHSH expression in terms of correlation functions, \( S \equiv |E_{11} + E_{12} - E_{21} - E_{22}| \leq 2 \) with \( E_{ij} = 2p(A = B|a_i,b_j) - 1 \), yields \( S = 2(-C - 1) = 2.425 \) for run 1 and \( S = 2.502 \) for run 2, violating the corresponding local-realist bound of 2.

### Predictability of Settings

We need to take into account two sources of imperfections in the experiment that can lead to an excess predictability of the setting choices. The excess predictability \( \epsilon \) quantifies the fraction of runs in which — given all possible knowledge about the setting generation process that can be available at the emission event of the particle pairs and thus at the distant measurement events — one could predict a specific setting better than what would simply be inferred from the overall bias of the setting choices. Loosely speaking, \( \epsilon \) quantifies the fraction of runs in which the locality and freedom-of-choice assumptions fail.
given by the ratio of noise rate to total rate, \( n_{a_1}/r_{a_1} = 0.017 \) \( n_{a_2}/r_{a_2} = 0.034 \) for run 1. Similarly, the noise contribution to the predictability for \( b_1 \) (\( b_2 \)) is given by \( n_{b_1}/r_{b_1} = 0.028 \) \( n_{b_2}/r_{b_2} = 0.011 \) for run 1.

The second source of imperfection is that a certain fraction of stellar photons leaves the dichroic mirror in the wrong output port. We index the wrong-way fractions \( f_{j'-i} \) as defined in Table III with \( i' \rightarrow i \) denoting either \( 1 \rightarrow 2 \) or \( 2 \rightarrow 1 \).

With \( (A) \) and \( (B) \) denoting Alice and Bob, we can write

\[
\begin{align*}
    r_{a_1} &= \left( 1 - f_{j' \rightarrow i}^{(A)} \right) s_j^{(A)} + f_{j' \rightarrow i}^{(A)} s_{j'}^{(A)} + n_{a_1}, \\
    r_{b_j} &= \left( 1 - f_{j' \rightarrow j}^{(B)} \right) s_j^{(B)} + f_{j' \rightarrow j}^{(B)} s_{j'}^{(B)} + n_{b_j}.
\end{align*}
\]

Here \( s_j^{(A)} (s_j^{(B)}) \) is the detected rate of stellar photons at Alice (Bob) which have a color that, when correctly identified, leads to the setting choice \( a_i \) (\( b_j \)). Each rate in Eq. (19) is a sum of three terms: correctly identified stellar photons, incorrectly identified stellar photons that should have been the other setting, and the noise rate. The four expressions in Eq. (19) allow us to find the four rates \( s_j^{(A)} \) and \( s_j^{(B)} \) as functions of the \( f \) parameters.

We now want to quantify the setting predictability due to the dichroic mirror errors. We imagine a hidden-variable model with arbitrary local power with the following restrictions: It cannot use non-detections to its advantage, and it can only alter at most certain fractions of the incoming stellar photons, which are quantified by the dichroic mirror error probabilities. We first focus only on Alice’s side. We assume that in a certain fraction of runs the local-realist model ‘attacks’ by enforcing a specific setting value and choosing hidden variables that optimize the measurement results to maximize the Bell violation. This could for instance happen with a hidden (slower than light) signal from the source to Alice’s dichroic mirror. Let us assume that \( q_{a_1} \) is the fraction of runs in which the model decides to generate setting \( a_1 \). If the incoming stellar photon would, under correct identification, have led to setting \( a_2 \), this ‘overruling’ gets reflected in the dichroic mirror error probability \( f_{j' \rightarrow i}^{(A)} \). In fact, we can equate \( q_{a_1} = f_{j' \rightarrow i}^{(A)} \), as the commitment to enforce setting \( a_1 \) to occur, independent of knowledge of the incoming photon’s wavelength. Thus, the probability to enforce \( a_1, q_{a_1} \), is identical to the conditional probability \( f_{j' \rightarrow i}^{(A)} \) that \( a_1 \) is enforced although \( a_2 \) would have been generated otherwise. The predictability from this ‘overruling’ is quantified by \( f_{j' \rightarrow i}^{(A)} s_i^{(A)}/r_{a_1} \), i.e. the fraction of \( a_1 \) settings which stem from stellar photons that should have led to setting \( a_2 \).

On the other hand, if the incoming stellar photon would have led to setting \( a_1 \) anyway, there is no visible ‘overruling’ and the attack remains hidden, while the model still produces outcomes that maximize the Bell violation. The predictability from this is quantified by

\[
\epsilon_{ij} = \epsilon_{a_1} + \epsilon_{b_j}.
\]
For both runs, Eqs. (22) and (26) yield $\epsilon \pm \sigma_\epsilon = \epsilon_{ij} \pm \sigma_{\epsilon_{ij}}$. If this number is larger than 1, $\epsilon_{ij}$ is set to 1.

We conservatively assume that all predictable events are maximally exploited by a local hidden-variable model. Then, in fact, the largest of the four fractions, i.e.,

$$\epsilon \equiv \max_j \epsilon_{ij} \equiv \max_i \epsilon_{ai} + \max_j \epsilon_{bj},$$

(22)

can be reached for the CHSH expression $C$.

To make this clear, let us consider the simple hidden-variable model in which the outcome values are always $A_1 = -1, A_2 = +1, B_1 = +1, B_2 = +1$, with subscripts indicating the respective setting. The first two probabilities in Eq. (18) are each 0 (only anticorrelations), the last two are each 1 (only correlations), and $C = 0$. Now, if in a fraction $\epsilon_{ai}$ of all coincidence events with setting combination $a_i b_i$ there is setting information of one party available at the source or the distant measurement event, then that latter measurement outcome can be “re-programmed” to produce an anticorrelation. Hence, we have $p(A = B(a_2 b_1)) = 0$ in that $\epsilon_{ai}$ subensemble, and $p(A = B(a_2 b_1)) = 1 - \epsilon_{ai}$ in total. This leads to $C = \epsilon_{ai}$.

Similar examples can be constructed for the other fractions. The predictabilities $\epsilon_{ij}$ thus require us to adapt the CHSH inequality of Eq. (18) to (see Ref. [9])

$$C \leq \epsilon.$$

(23)

The dichroic mirror errors were characterized, taking into account the spectra of the stars and all optical elements. Using the values for $f_{i',f}$ in Table III and the total and noise rates from Table V yields a predictability of $\epsilon = 0.1779$ for run 1, such that our observed value $C = 0.2125$ still represents a violation of the adapted inequality of Eq. (23). Likewise for run 2, we find $\epsilon = 0.1609$, again yielding $C = 0.2509 > \epsilon$. See Table V.

### Uncertainty on the Settings Predictability

We temporarily drop the labels for Alice and Bob. Assuming that the rates $r_i, n_i$, and the $f$ parameters are independent (which follows from our assumption of fair sampling for all detected photons), error propagation of

Eq. (20) yields an uncertainty estimate for $\epsilon_i$ given by

$$\sigma_{\epsilon_i}^2 = r_i^{-4} \left( r_i^2 \sigma_{\epsilon_{ij}}^2 + \left( 1 - f_r \right) \sigma_{\epsilon_r}^2 + f_r \sigma_{\epsilon_n}^2 \right) + \left[ n_i \left( 1 - f_{\epsilon_{ij}} \right) + f_r \epsilon_r \left( r_i - n_i \right) \right]^2 \sigma_{\epsilon_r}^2,$$

(24)

where we note that $s = r_1 + r_2 - n_1 - n_2$. Eq. (24) holds for Alice or Bob by applying appropriate labels. If we assume Alice and Bob’s predictability contributions from Eq. (21) are independent, we find

$$\sigma_{\epsilon_{ij}} = \sqrt{\sigma_{\epsilon_{ai}}^2 + \sigma_{\epsilon_{bj}}^2}.$$

(25)

with an estimated uncertainty on $\epsilon$ from Eq. (22) of

$$\sigma_{\epsilon} = \sqrt{\sigma_{\epsilon_{max,ai}}^2 + \sigma_{\epsilon_{max,bj}}^2}.$$

(26)

where $\sigma_{\epsilon_{max,ai}}$ is the uncertainty from Eq. (24) on the term which maximizes $\epsilon_{ai}$, and likewise for Bob. For both runs 1 and 2, assuming values and errors on the total and noise rates from Table V wrong-way fractions $f$ from Table III with conservative fractional errors of $\sigma_i/f = 0.1$, Table V shows values of $\epsilon_{ij}$ from Eqs. (22-24), $\sigma_{\epsilon_{ij}}$ from Eqs. (24-25), and $\sigma_\epsilon$ from Eq. (26).

### Statistical significance

There exist several different ways to estimate the statistical significance for experimental runs 1 and 2. The result of any such statistical analysis is a $p$-value, i.e., a bound for the probability that the null hypothesis — local realism with $\epsilon$ predictability, biased detector-setting frequencies, fair sampling, fair coincidences, and any other additional assumptions — could have produced the experimentally observed data by a random variation.

Until recently, it was typical in the literature on such Bell tests to estimate a $p$-value under several assumptions (e.g., [10]): that each trial was independent and identically distributed (i.i.d.), and that the hidden-variable mechanism could not make any use of “memory” of the settings and outcomes of previous trials. Under those assumptions, one typically applied Poisson statistics for single coincidence counts, and assumed that the underlying statistical distribution was Gaussian. Moreover, it was typical to neglect the excess predictability, $\epsilon$. Applied to our experimental data, such methods yield what we consider to be overly optimistic estimates, suggesting violation of the CHSH inequality by $\nu \geq 39.8$ and 42.7 standard deviations for runs 1 and 2, respectively.

However, such an approach assumes that the measured coincidence counts $N_{ij}^{AB}$ are equal to their expected values, but then contradicts this assumption by calculating the probability that the $N_{ij}^{AB}$ could have values differing

<table>
<thead>
<tr>
<th>Run</th>
<th>$\epsilon_{11} \pm \sigma_{\epsilon_{11}}$</th>
<th>$\epsilon_{12} \pm \sigma_{\epsilon_{12}}$</th>
<th>$\epsilon_{21} \pm \sigma_{\epsilon_{21}}$</th>
<th>$\epsilon_{22} \pm \sigma_{\epsilon_{22}}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0.13521 ± 0.07645</td>
<td>0.17791 ± 0.11915</td>
<td>0.11915</td>
<td>0.16094 ± 0.10533</td>
</tr>
<tr>
<td>2</td>
<td>0.10533 ± 0.08917</td>
<td>0.16094 ± 0.14477</td>
<td>0.08917</td>
<td>0.16094 ± 0.11915</td>
</tr>
</tbody>
</table>

Table V. For runs 1 and 2, we compute $\epsilon_{ij}$ with Eq. (21) and $\sigma_{\epsilon_{ij}}$ with Eqs. (24-25). We use values and errors on the total and noise rates from Table IV along with 10% fractional uncertainties on the dichroic mirror wrong-way fractions in Table III. For both runs, Eqs. (22) and (26) yield $\epsilon \pm \sigma_\epsilon = \epsilon_{ij} \pm \sigma_{\epsilon_{ij}}$. 
by several standard deviations from their expected values. Plus, as recent work has emphasized (e.g., [9]), excess predictability $\epsilon$ must be taken into account when estimating statistical significance for any violations of the CHSH inequality.

More recently, several authors have produced improved methods for calculating $p$-values for Bell tests. These newer approaches do not assume i.i.d. trials, and also, more conservatively, allow the hidden-variable model to exploit “memory” of previous settings and outcomes. Whereas the “memory loophole” cannot achieve Bell violation, incorporating possible memory effects does require modified calculations of statistical significance [9][11][13].

Although these new works represent a clear advance in the literature, unfortunately they are not optimized for use with our particular experiment. For example, the unequal settings probabilities (bias) for our experiment limit the utility of the bounds derived in [13][14], as the resulting $p$-values are close to 1. Likewise, one may follow the approach of [9][11][12] and use the Hoeffding inequality [15]. However, it is known in general that such bounds routinely overestimate $p$ — and hence underestimate the genuine statistical significance of a given experiment — by a substantial amount (see, e.g., [14]).

Therefore, in this section we present an ab initio calculation of the $p$-value tailored more specifically to our experiment. This method yields what we consider to be reasonable upper bounds on the $p$-values, which are still highly significant even with what we regard as a conservative set of assumptions. Our calculation incorporates predictability of settings and allows the local-realist hidden-variable theory to exploit memory of previous detector settings and measurement outcomes. We present essential steps in the calculation here, and defer fuller discussion to future work.

We consider a quantity $W$, which is a weighted measure of the number of “wins,” that is, the number of measurement outcomes that contribute positively to the CHSH quantity $C$, defined in Eq. (18). A win consists of $A = B$ for settings pair $a_i b_j$, and $A \neq B$ for any other combination of settings. Thus we define $N_{ij}^{\text{win}} \equiv N_{1i}^{A=B} N_{1j}^{A=B} N_{2i}^{A=B} N_{2j}^{A=B}$, and

$$W = \sum_{ij} \frac{N_{ij}^{\text{win}}}{q_{ij}(1 - \epsilon_{ij})},$$

(27)

where $q_{ij} \equiv N_{ij}/N$ is the fraction of trials in which settings combination $ij$ occurs, and $\epsilon_{ij}$, defined in Eq. (27), is the probability that a given trial will be “corrupt.” A trial is considered “corrupt” if it (1) involved a noise (rather than stellar) photon, or (2) involved a dichroic mirror error, or (3) was previewed by the hidden-variable theory for the purpose of considering a dichroic mirror error, but was passed over because the stellar photon already had the desired color. The occurrence of a corruption in any trial is taken to be an independent random event, which has probability $\epsilon_{ij}$ that depends on the settings pair $a_i b_j$. We assume that for “uncorrupt” trials, the hidden-variable theory has no information about what the settings pair will be beyond the probabilities $q_{ij}$.

We assume that the hidden-variable theory can exploit each corrupted trial and turn it into a win. We further assume that the occurrence of these corrupt events cannot be influenced by either the experimenter or the hidden-variable theory; they occur with uniform probability $\epsilon_{ij}$ in each trial. We consider the probabilities $\epsilon_{ij}$ to be known (to within some uncertainty $\sigma_{\epsilon_{ij}}$), but the actual number of corrupt trials to be subject to statistical fluctuations.

The $p$-value is the probability that a local-realist hidden-variable theory, using its best possible strategy, could obtain a value of $W$ as large as the observed value. To define this precisely, we must be clear about the ensemble that we are using to define probabilities. It is common to attempt to describe the ensemble of all experiments with the same physical setup and the same number of trials. Yet it is difficult to do this in a precise way, because one has to use the statistics of settings choices observed in the experiment to determine the probabilities for the various settings. From a Bayesian point of view, this requires the assumption of a prior probability distribution on settings probabilities, and the answers one finds for $p$ would depend on what priors one assumes.

We avoid such issues by considering the actual number $N_{ij}$ of the occurrences of each settings choice $a_i b_j$ as given. The relevant ensemble is then the ensemble of all possible orders in which the settings choices could have occurred. The $p$-value will then be the fraction of orders for which the hidden-variable theory, using its best strategy, could obtain a value of $W$ greater than or equal to the value obtained in the experiment.

We may motivate the form of $W$ in Eq. (27) as follows. In the absence of noise or errors, the hidden-variable model could specify which outcomes $(A, B)$ will arise for each of the possible settings $(i, j)$. The best plans will win for three of the four possible settings pairs, but will lose for one of the possible settings pairs. Hence a plan may be fully specified by identifying which settings pair will be the loser. (There will actually be two detailed plans for such a specification, related by a reversal of all outcomes, but we may treat such plans as equivalent.)

In the presence of noise and errors, for each time the settings pair is $a_i b_j$, there is a probability $\epsilon_{ij}$ that the trial is corrupt. If the trial is corrupt, it automatically registers as a win. If it is not corrupt, then it has a probability $P_{ij}^{\text{win}}$ of registering as a win, where we take $P_{ij}^{\text{win}}$ to be $p(A = B | a_i b_j)$ for $(ij) = (22)$, and $p(A \neq B | a_i b_j)$ for the
other three cases. Then we may write
\[ \langle N_{ij}^{\text{win}} \rangle = \left[ \epsilon_j + (1 - \epsilon_j) P_{ij}^{\text{win}} \right] N_{ij}, \] (28)
which may be solved for \( P_{ij}^{\text{win}} \):
\[ \alpha_{ij}^{\text{win}} = \frac{\langle N_{ij}^{\text{win}} \rangle}{N_{ij}(1 - \epsilon_j)} - \epsilon_j. \] (29)

The CHSH inequality may be written \( \sum_{ij} \alpha_{ij}^{\text{win}} \leq 3 \), so Eq. (29) implies that
\[ \sum_{ij} \alpha_{ij}^{\text{win}} \leq (3 + \tilde{\epsilon}) N, \] (30)
where we have defined
\[ \tilde{\epsilon} = \sum_{ij} \frac{\epsilon_j}{1 - \epsilon_j}. \] (31)
The lefthand side of Eq. (30) motivates our ansatz for \( W \) in Eq. (27).

The function \( W \), which is a random variable, may be expressed in terms of a set of more elementary random variables. We label the trials by \( \alpha \), so for each trial \( \alpha \) there will be a set of random variables:
\[ F_{ij}^{\alpha} = \begin{cases} 1 & \text{if the settings pair is } a_i b_j \text{ in trial } \alpha \\ 0 & \text{otherwise} \end{cases}, \]
\[ G^{\alpha} = \begin{cases} 1 & \text{if the trial } \alpha \text{ is corrupt} \\ 0 & \text{otherwise} \end{cases}, \]
\[ U^{\alpha} = \begin{cases} 1 & \text{if the trial } \alpha \text{ is uncorrupt} \\ 0 & \text{otherwise} \end{cases}, \] (32)
with \( G^{\alpha} + U^{\alpha} = 1 \). We also define the functions
\[ \omega_{ij}^{\alpha} = \begin{cases} 1 & \text{if the settings pair } a_i b_j \text{ in trial } \alpha \text{ is a win} \\ 0 & \text{otherwise} \end{cases}, \]
\[ \tilde{\omega}_{ij}^{\alpha} = \begin{cases} 1 & \text{if the settings pair } a_i b_j \text{ in trial } \alpha \text{ is a loss} \\ 0 & \text{otherwise} \end{cases}. \] (33)
with \( \omega_{ij}^{\alpha} + \tilde{\omega}_{ij}^{\alpha} = 1 \). Unlike the variables in Eq. (32), \( \omega_{ij}^{\alpha} \) and \( \tilde{\omega}_{ij}^{\alpha} \) are not random; they are under the control of the hidden-variable mechanism. The square of each of the quantities in Eqs. (32) and (33) is equal to itself, since their only possible values are 0 and 1.

Our goal is to evaluate \( \sigma_{W}^2 = \langle W^2 \rangle - \langle W \rangle^2 \). We begin by calculating \( \langle W \rangle = \sum_\alpha \langle W^\alpha \rangle \). In terms of the quantities in Eqs. (32) and (33), we may write
\[ W^{\alpha} = \sum_{ij} F_{ij}^{\alpha} \frac{U^{\alpha} \omega_{ij}^{\alpha} + G^{\alpha}}{q_{ij}(1 - \epsilon_j)}. \] (34)

Since the settings are chosen randomly on each trial, we assume that all orderings of the setting choices are equally likely, and are independent of the occurrence of corruptions. This implies that \( \langle F_{ij}^{\alpha} U^{\alpha} \rangle = q_{ij}(1 - \epsilon_j) \) and \( \langle F_{ij}^{\alpha} G^{\alpha} \rangle = q_{ij}\epsilon_j \), independent of \( \alpha \). Then we find
\[ \langle W^{\alpha} \rangle = \sum_{ij} q_{ij} \left[ (1 - \epsilon_j) \omega_{ij}^{\alpha} + \epsilon_j \right] \]
\[ = \sum_{ij} \omega_{ij}^{\alpha} + \sum_{ij} \frac{\epsilon_j}{1 - \epsilon_j} = 3 + \tilde{\epsilon}, \] (35)
and hence
\[ \langle W \rangle = N(3 + \tilde{\epsilon}). \] (36)

To evaluate \( \langle W^2 \rangle \) we write
\[ \langle W^2 \rangle = \sum_\alpha \sum_\beta \langle W^\alpha W^\beta \rangle = \sum_\alpha \langle W^2 \rangle + \sum_\alpha \sum_{\beta \neq \alpha} \langle W^\alpha W^\beta \rangle. \] (37)

For the first term, we have
\[ \langle W^2 \rangle = \sum_{ij} \sum_{k\ell} \frac{\langle F_{ij}^{\alpha} F_{k\ell}^{\beta} (U^{\alpha} \omega_{ij}^{\alpha} + G^{\alpha})(U^{\beta} \omega_{k\ell}^{\beta} + G^{\beta}) \rangle}{q_{ij}q_{k\ell}(1 - \epsilon_j)(1 - \epsilon_{k\ell})} \]
\[ = \sum_{ij} \frac{\omega_{ij}^{\alpha}}{q_{ij}(1 - \epsilon_j)} + \sum_{ij} \frac{\epsilon_j}{q_{ij}(1 - \epsilon_j)^2}, \] (38)
where the second line follows upon noting that \( F_{ij}^{\alpha} F_{k\ell}^{\beta} = 0 \) if \((ij) \neq (k\ell)\), and using \( (F_{ij}^{\alpha})^2 = F_{ij}^{\alpha} \), \( (U^{\alpha})^2 = U^{\alpha} \). We therefore find
\[ \sum_\alpha \langle W^2 \rangle = N \left[ \sum_{ij} \frac{1 - f_{ij}}{q_{ij}(1 - \epsilon_j)} + \sum_{ij} \frac{\epsilon_j}{q_{ij}(1 - \epsilon_j)^2} \right]. \] (39)

where we have defined \( f_{ij} \) as the fraction of trials for which the hidden-variable theory chooses \((ij)\) to be the losing settings pair.

For the second term on the righthand side of Eq. (37), we have
\[ \sum_\alpha \sum_{\beta \neq \alpha} \langle W^\alpha W^\beta \rangle = \sum_\alpha \sum_{\beta \neq \alpha} \sum_{ij} \sum_{k\ell} \frac{\langle F_{ij}^{\alpha} F_{k\ell}^{\beta} (U^{\alpha} \omega_{ij}^{\alpha} + G^{\alpha})(U^{\beta} \omega_{k\ell}^{\beta} + G^{\beta}) \rangle}{q_{ij}q_{k\ell}(1 - \epsilon_j)(1 - \epsilon_{k\ell})} \]
\[ = \sum_{ij} \sum_{k\ell} \sum_\alpha \sum_{\beta \neq \alpha} \frac{q_{ij}q_{k\ell}(Nq_{k\ell} - \delta_{ij,k\ell})}{N - 1} \times \left[ (1 - \epsilon_j)\omega_{ij}^{\alpha} + \epsilon_j \right] \left[ (1 - \epsilon_{k\ell})\omega_{k\ell}^{\beta} + \epsilon_{k\ell} \right] \]
\[ = T_1 + T_2, \] (40)
where \( \delta_{ij,k\ell} = 1 \) if \((ij) = (k\ell)\) and 0 otherwise. (We have used the fact that for each of the \( N_{ij} \) values of \( \alpha \) for
which $F_{ij}^\beta = 1$, there are $N_{ij} - 1$ values of $\beta \neq \alpha$ for which $F_{ij}^\beta = 1$.

To further simplify Eq. (40), we first assume that the hidden-variable theory cannot exploit memory of previous settings or outcomes. In that case, we may neglect correlations between $F_{ij}^\alpha$ and $\omega_{ij}^\beta$, and perform a full ensemble average. (We will relax this assumption below.) Proceeding as above, the term $T_1$ may then be rewritten

$$T_1 = \frac{N}{N - 1} \sum_{ij} \sum_{k \neq \ell} \sum_\alpha \sum_{\beta, \alpha} \frac{1}{(1 - \epsilon_{ij})(1 - \epsilon_{\ell \ell})} \times \left[ (1 - \epsilon_{ij})\omega_{ij}^\alpha + \epsilon_{ij} \right] \left[ (1 - \epsilon_{\ell \ell})\omega_{\ell \ell}^\beta + \epsilon_{\ell \ell} \right]$$

$$= N^2(3 + \bar{\epsilon})^2,$$

where we have made use of the fact that $\sum_{ij} 1/(1 - \epsilon_{ij}) = \sum_{ij}(1 - \epsilon_{ij})/(1 - \epsilon_{ij}) + \sum_{ij} \epsilon_{ij}/(1 - \epsilon_{ij}) = 4 + \bar{\epsilon}$. For the term $T_2$, we note that

$$\sum_\alpha \sum_\beta \omega_{ij}^\alpha \omega_{\ell \ell}^\beta = N(1 - f_{ij}) \left[ N(1 - f_{ij}) - 1 \right].$$

Then $T_2$ may be rewritten

$$T_2 = -\frac{1}{N - 1} \sum_{ij} \sum_{\alpha} \sum_{\beta} \frac{1}{q_{ij}(1 - \epsilon_{ij})^2} \times \left[ (1 - \epsilon_{ij})\omega_{ij}^\alpha + \epsilon_{ij} \right] \left[ (1 - \epsilon_{ij})\omega_{ij}^\beta + \epsilon_{ij} \right]$$

$$= -\frac{N}{N - 1} \sum_{ij} \frac{(1 - f_{ij})}{q_{ij}} \left( N(1 - f_{ij}) - 1 \right) + N \sum_{ij} \left\{ \frac{2f_{ij}(1 - f_{ij})}{q_{ij}(1 - \epsilon_{ij})} + \frac{\epsilon_{ij}^2}{q_{ij}(1 - \epsilon_{ij})^2} \right\}.$$

Following some straightforward algebra, Eqs. (39), (41), and (43) yield

$$\sigma_w^2 = \frac{N^2}{N - 1} \sum_{ij} f_{ij}(1 - f_{ij})q_{ij} + N \sum_{ij} f_{ij}\epsilon_{ij}.$$  \hspace{1cm} (44)

The $f_{ij}$ are under the control of the hidden-variable theory, so we make the conservative assumption that the hidden-variable theory may choose the $f_{ij}$ so as to maximize $\sigma_w$. To impose the constraint that $\sum_{ij} f_{ij} = 1$, we introduce a Lagrange multiplier $\lambda$:

$$L = \frac{N^2}{N - 1} \sum_{ij} f_{ij}(1 - f_{ij})q_{ij} + N \sum_{ij} f_{ij}\epsilon_{ij}$$

$$+ \lambda \left( \sum_{ij} f_{ij} - 1 \right).$$

Setting $\partial L/\partial f_{ij} = 0$, we find the optimum values $f_{ij}^{opt}$. By inserting these into the normalization condition $\sum_{ij} f_{ij}^{opt} = 1$, we may solve for $\lambda$, which in turn yields

$$f_{ij}^{opt} = \frac{1}{2} - q_{ij} + \frac{N - 1}{2N} \left[ \frac{\epsilon_{ij}}{1 - \epsilon_{ij}} - \bar{\epsilon}q_{ij} \right].$$  \hspace{1cm} (46)

Inserting $f_{ij}^{opt}$ into Eq. (44) for $\sigma_w^{opt}$, we find

$$\left(\sigma_w^{opt}\right)^2 = \frac{N^2}{4(N - 1)} \left( \sum_{ij} \frac{1}{q_{ij}} \right) - N\bar{\epsilon}$$

$$+ \frac{N}{4} \sum_{ij} \frac{\epsilon_{ij}}{q_{ij}(1 - \epsilon_{ij})}$$

$$- \frac{1}{4}(N - 1)\bar{\epsilon}^2 + \frac{1}{4} \sum_{ij} \frac{(N - \epsilon_{ij})\epsilon_{ij}}{q_{ij}(1 - \epsilon_{ij})^2}.$$  \hspace{1cm} (47)

For run 1, Eq. (46) yields an unphysical $f_{12} < 0$ for our data. Upon employing a second Lagrange multiplier to ensure both that $\sum_{ij} f_{ij} = 1$ and $f_{ij} \geq 0$, we find

$$f_{ij}^{opt} = \frac{1}{2} - q_{ij} + \frac{N - 1}{2N} \left[ \frac{\epsilon_{ij}}{1 - \epsilon_{ij}} - q_{ij} \left( \frac{N}{N - 1} + \bar{\epsilon} - \frac{\epsilon_{ij}}{1 - \epsilon_{ij}} \right) \right],$$

such that $f_{11}^{opt} = 0.376$, $f_{12}^{opt} = 0$, $f_{21}^{opt} = 0.483$, and $f_{22}^{opt} = 0.141$. For run 1, one must substitute Eq. (48) into Eq. (44) in order to find $\sigma_w^{opt}$. For run 2, on the other hand, Eq. (46) yields $f_{ij} > 0 \forall (ij)$, with $f_{11}^{opt} = 0.101$, $f_{12}^{opt} = 0.062$, $f_{21}^{opt} = 0.428$, and $f_{22}^{opt} = 0.409$, allowing $\sigma_w^{opt}$ to be computed with Eq. (47).

Using the values for total and noise rates ($r, n$) in Table IV, dichroic mirror wrong-way fractions ($f$) in Table III, values of $\eta_{ij}$ inferred from Eqs. (11), and 14) and the probabilities for corrupt trials $\epsilon_{ij}$ in Table V, values for $W, \langle W \rangle$, and $\sigma_w^{opt}$ for both runs are listed in Table VI.

<table>
<thead>
<tr>
<th>Run</th>
<th>$W$</th>
<th>$\langle W \rangle$</th>
<th>$\sigma_w^{opt}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>5.0249 $\times 10^4$</td>
<td>4.8954 $\times 10^3$</td>
<td>954.3</td>
</tr>
<tr>
<td>2</td>
<td>3.3030 $\times 10^3$</td>
<td>3.1754 $\times 10^3$</td>
<td>682.6</td>
</tr>
</tbody>
</table>

**TABLE VI.** For runs 1 and 2, values for $W$ and $\langle W \rangle$ from Eqs. (27) and (36) are shown, as well as values of $\sigma_w^{opt}$ from Eqs. (44) and (48) for run 1 and Eq. (47) for run 2.

Next we take into account the uncertainty in the predictabilities $\epsilon_{in}$ and $\epsilon_{in}$. The quantity of interest is

$$\bar{\eta} = \frac{W - \langle W \rangle}{\sigma_w^{opt}}.$$  \hspace{1cm} (49)

The quantities $W, \langle W \rangle$, and $\sigma_w^{opt}$ all depend on $\epsilon_{in}$ and $\epsilon_{in}$, along with the $N_{ij}$ values, which are taken as given. Therefore, we only need to propagate uncertainties on $\epsilon_{in}$ and $\epsilon_{in}$ to compute the uncertainty on $\bar{\eta}$, which we denote $\Delta_\bar{\eta}$.

We assume no covariance between $\epsilon_{in}$ and $\epsilon_{in}$. This again follows from our assumptions of independence for
Alice and Bob as well as fair sampling for all detected photons, which implies \( r_i, n_i, \) and \( f_i \rightarrow \) (the inputs to \( e_{ni} \) and \( e_{ji} \)) are independent. An estimate for \( \Delta_\nu \) is then given by:

\[
\Delta_\nu^2 = \sum_{ni} \left( \frac{\partial \nu}{\partial e_{ni}} \right)^2 \sigma_{e_{ni}}^2 + \sum_{bj} \left( \frac{\partial \nu}{\partial e_{bj}} \right)^2 \sigma_{e_{bj}}^2
\]

\[
= \sum_{ni} \left( \frac{\sigma_{e_{ni}}}{\sigma_W} \right)^2 \left( \sum_{j} q_{ij}(1 - e_{ij})^2 \right) + \sum_{bj} \left( \frac{\sigma_{e_{bj}}}{\sigma_W} \right)^2 \left( \sum_{i} q_{ij}(1 - e_{ij})^2 \right),
\]

where

\[
E_{ij} \equiv N_{ij}^{\text{win}} - N_{ij}^{\text{lose}} - \left( \frac{\bar{\nu} N}{2\sigma_W^2} \right) f_{ij}^{\text{opt}},
\]

and we recall from Eq. (21) that \( e_{ij} = e_{ni} + e_{bj}. \) We may now compute how \( \sigma_{e_{ni}} \) and \( \sigma_{e_{bj}} \) affect the statistical significance of each run. The naive number of standard deviations \( \bar{\nu} \) in Eq. (49) implicitly assumed \( \sigma_{e_{ni}}, \sigma_{e_{bj}} = 0, \) and therefore \( \Delta_\nu = 0. \) If we allow for an uncertainty in \( \nu \) equal to \( n \) times the 1-\( \sigma \) uncertainty in \( \nu, \) then we should calculate the \( p \)-value using

\[
\nu_n \equiv \bar{\nu} - n\Delta_\nu.
\]

If we choose \( n \) so that \( n = \nu_n, \) then

\[
\nu_n = \frac{\bar{\nu}}{1 + \Delta_\nu}.
\]

Assuming a Gaussian distribution for large-sample experiments, we conclude that the conditional probability that the hidden variable mechanism could achieve a value of \( W \) as large as the observed value \( W_{\text{obs}}, \) assuming that the true value of \( \nu \geq \nu_n, \) is given by \( p_{\text{cond}} = \frac{1}{2}\text{erfc} (\nu_n / \sqrt{2}). \) Since we chose \( n = \nu_n, \) if we assume Gaussian statistics for the uncertainty in \( \nu, \) then there is an equal probability that the true value of \( \nu \) is less than \( \nu_n, \) in which case our analysis does not apply, and we must conservatively assume that \( W \) might exceed \( W_{\text{obs}}. \) Thus, the \( p \)-value corresponding to the total probability that \( W \geq W_{\text{obs}} \) is bounded by \( p = 2p_{\text{cond}}. \) Again assuming Gaussian statistics, we can relate \( p \) to an equivalent \( \nu, \) by \( p = \frac{1}{2}\text{erfc}(\nu / \sqrt{2}). \) Proceeding in this way, we find the values for \( \bar{\nu}, \Delta_\nu, \nu, \) and \( p \) listed in Table VII.

<table>
<thead>
<tr>
<th>Run</th>
<th>( \bar{\nu} )</th>
<th>( \Delta_\nu )</th>
<th>( \nu )</th>
<th>( p )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>13.57</td>
<td>0.79905</td>
<td>7.54</td>
<td>4.64 \times 10^{-14}</td>
</tr>
<tr>
<td>2</td>
<td>18.71</td>
<td>0.53999</td>
<td>12.15</td>
<td>5.93 \times 10^{-28}</td>
</tr>
</tbody>
</table>

TABLE VII. Values for \( \bar{\nu}, \Delta_\nu, \nu, \) and \( p \) for runs 1 and 2.

Memory of Previous Trials

Next we consider possible memory effects. We define the quantity \( \bar{W} \equiv W - (3 + \bar{\nu})/N. \) Then \( \bar{W} = 0, \) regardless of what plan the hidden-variable theory uses. On the other hand, the hidden-variable theory can affect the standard deviation of \( W. \) If we denote by \( W_0 \) the value of \( W \) obtained in the experiment, then the \( p \)-value we seek is the probability that the hidden-variable theory could have achieved \( \bar{W} \geq \bar{W}_0 \) by chance. To discuss an experiment in progress, we define

\[
\bar{W}_n \equiv \sum_{a=1}^n (W_a - 3 - \bar{\nu})/N,
\]

which is the contribution to \( \bar{W} \) after \( n \) trials.

For sufficiently large \( N, \) we may assume that the probabilities are well approximated by a Gaussian probability distribution. Then we expect that as long as \( \bar{W}_n \leq \bar{W}_0, \) the best strategy for the hidden-variable theory is to maximize \( \sigma_{\bar{\nu}} \), so that the number of standard deviations to its goal is as small as possible. When and if \( \bar{W}_n \) passes \( \bar{W}_0, \) on the other hand, then its best strategy is to minimize \( \sigma_{\bar{\nu}} \), so as to minimize the probability that \( \bar{W} \) might backslide to \( \bar{W} \leq \bar{W}_0. \)

We define \( N_{\text{lose,ij}} \) as the number of trials for which the hidden-variable theory selects settings \((ij)\) as the loser. Then we seek to estimate \( p_{\text{left}}(\nu|N_{\text{lose,ij}}) \equiv p(\bar{W}_n < 0), \) under the assumption that the hidden-variable loser selection is given by \( N_{\text{lose,ij}}. \) That is, \( p_{\text{left}}(\nu|N_{\text{lose,ij}}) \) is the probability that after \( n \) trials, the net change in \( W \) has been to the left (i.e., negative). For large \( n, \) we expect the probability distribution for \( W \) to become a Gaussian with zero mean, so that \( p_{\text{left}}(\nu|N_{\text{lose,ij}}) \) should approach 1/2, for any hidden-variable theory loser selection. For smaller \( n, \) however, \( p_{\text{left}}(\nu|N_{\text{lose,ij}}) \) can reach some maximum value \( B > 1/2. \)

Finally, we define \( p_1 \) to be the probability that \( \bar{W}_n \geq \bar{W}_0 \) for some \( n \) in the range \( 1 \leq n \leq N, \) under the assumption that the hidden-variable theory consistently makes choices that maximize \( \sigma_{\bar{\nu}}. \) Consider some particular sequence of trials that contributes to \( p_1, \) that is, a sequence for which \( \bar{W}_n \geq \bar{W}_0 \) for some \( n. \) The continuation of this sequence for the rest of the experiment (assuming that the hidden-variable theory continues to make choices that maximize \( \sigma_{\bar{\nu}} \)) can do one of two things: it can finish the experiment with \( \bar{W} \leq \bar{W}_0, \) or it can finish the experiment with \( \bar{W} < \bar{W}_0. \) In the first case, this sequence contributes to the \( p \) value we calculated in the previous subsection, whereas in the second case it does not. The second case is an instance of backsliding, for which we know that the probability is at most \( B. \) Hence the probability of the first case is at least \( 1 - B, \) so the \( p \) value we seek, \( p_{\text{mem}}, \) satisfies

\[
p_{\text{mem}} \leq \frac{p}{1 - B},
\]
Lastly, we check whether our data are consistent with the no-signaling principle. This principle demands that, under space-like separation, local outcome probabilities must not depend on the setting of the distant party:

\[
p(A = +|a,b_j) = p(A = +|a,b_j),
\]
\[
p(B = +|a,b_j) = p(B = +|a,b_j).
\]

The analogous equations for the ‘−’ outcomes follow trivially from \(p(A = −|a,b_j) = 1 - p(A = +|a,b_j)\). Let us denote by \(N^{+}_{a_i} (N^{-}\_{b_j})\) the number Alice’s (Bob’s) outcomes ‘+’ where she (he) had setting \(a_i (b_j)\). The recorded data for experimental run 1, post-selecting only onto a valid setting choice (i.e., the click in the setting reader occurred within the time-interval \(\tau_{\text{used}}\) and not onto a coincident outcome at the distant location, were

\[
\begin{align*}
N^{+}_{a_1} & = 163292 & 550046 & N^{+}_{b_1} & = 562351 & 352896 \\
N^{+}_{a_2} & = 101289 & 340045 & N^{+}_{b_2} & = 2033046 & 1279635 \\
N^{-}_{a_1} & = 165593 & 555034 & N^{-}_{b_1} & = 480738 & 302277 \\
N^{-}_{a_2} & = 100848 & 340890 & N^{-}_{b_2} & = 1553010 & 976740
\end{align*}
\]

The data in Eq. (57) were obtained after applying the \(\tau_{\text{cut}}\) filter. We denote by \(N^{+}_{a_i,b_j} (N^{-}_{a_i,b_j})\) the value of \(N^{+}_{a_i} (N^{-}_{b_j})\) in the above table for distant setting \(b_j (a_i)\). A point estimate for \(p(A = +|a,b_j)\) is then given by \(N^{+}_{a_i,b_j}/(N^{+}_{a_i,b_j} + N^{-}_{a_i,b_j})\), and a point estimate for \(p(B = +|a,b_j)\) is given by \(N^{+}_{b_i,a_j}/(N^{+}_{b_i,a_j} + N^{-}_{b_i,a_j})\).

Under space-like separation of all relevant events, no-signaling must be obeyed in both local realism and quantum mechanics, since its violation would contradict special relativity. (An experimental violation of no-signaling would require the settings of the distant laboratory to be available at the local measurement station via faster-than-light communication or due to a common cause in the past.) For run 1, point estimates yield the following probabilities:

\[
\begin{align*}
p(A = +|a_1,b_1) & = 0.4965, & p(A = +|a_1,b_2) & = 0.4977, \\
p(A = +|a_2,b_1) & = 0.5011, & p(A = +|a_2,b_2) & = 0.4994, \\
p(B = +|a_1,b_1) & = 0.5391, & p(B = +|a_1,b_2) & = 0.5386, \\
p(B = +|a_2,b_2) & = 0.5669, & p(B = +|a_2,b_2) & = 0.5671.
\end{align*}
\]

The null hypothesis of no-signaling demands that the two conditional probabilities in each line should be equal. In order to test for signaling, we perform a pooled two-proportion \(z\)-test. The probabilities that the observed data or worse are obtained under the null hypothesis are 0.211, 0.177, 0.532, 0.654, respectively. (For the stars used in run 2, we obtain the probabilities 0.434, 0.342, 0.737, 0.582, respectively.) As all probabilities are large,
our data are in agreement with the no-signaling assumption. (We performed the same test on our data for runs 1 and 2 prior to applying the $\tau_{\text{cut}}$ filter, and likewise found no statistical evidence to suggest signaling.)

We remark that when post-selecting on coincident outcome events, i.e. using the counts in Eq. \[10\], the condition $p(A = +|a_1b_1, B = +) = p(A = +|a_2b_2, B = +)$, where $B = +$ denotes that Bob had a definite outcome (whose value is ignored), is violated significantly in both experiments. This can be attributed to the fact that the two total detection efficiencies for outcomes ‘+’ and ‘−’, especially on Bob’s side, were not the same. Let us denote the total detection efficiencies of Alice (Bob) for outcome ± by $\eta_a^{(A)} (\eta_b^{(B)})$, including all losses in the source, the link, and the detectors themselves. A detailed quantum-mechanical model for the data of run 1 suggests that the ratios of these efficiencies were $R^{(A)} \equiv \eta_a^{(A)}/\eta_a^{(A)} = 1.00$ for Alice and $R^{(B)} \equiv \eta_b^{(B)}/\eta_b^{(B)} = 0.81$ for Bob. The difference in $R^{(A)}$ and $R^{(B)}$ can fully be understood on the basis of the known efficiency differences of the detectors used. One can correct the counts in Eq. \[10\] for these efficiencies by multiplying all ‘+’ counts of Alice (Bob) by $\sqrt{R^{(A)}} (\sqrt{R^{(B)}})$, and dividing all her (his) ‘−’ counts by $\sqrt{R^{(A)}} (\sqrt{R^{(B)}})$. The corrected counts show no sign of a violation of no-signaling. This is also true for the data from run 2.

We finally note that, due to the low total detection efficiencies, our experiment had to make the assumptions of fair sampling and fair coincidences. This implies that low or imbalanced detection efficiencies are not exploited by hidden-variable models.

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