Quantum interference enables constant-time quantum information processing

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Science, medicine and engineering demand efficient information processing. It is a long-standing goal to use quantum mechanics to significantly improve such computations^{[1](#page-10-0)}. The processing routinely involves examining data as a function of complementary variables, e.g., time and frequency. This is done by the Fourier transform approximations which accurately compute inputs of 2^n 2^n samples in $O(n2^n)$ steps $^{\text{\textbf{2.}}}$ In the quantum domain, an analogous process exists, namely a Fourier transform of quantum amplitudes^{[3](#page-10-2)}, which requires exponentially fewer $O(n \log n)$ quantum gates. Here, we report a quantum fractional Kravchuk-Fourier transform, a related process suited to finite string processing 24 . Unlike previous demonstrations^{[5,](#page-10-3)[6](#page-10-4)}, our architecture involves only one gate, resulting in constant-time pro-cessing of quantum information. The gate exploits a generalized Hong-Ou-Mandel effect^{[7](#page-10-5)}, the basis for quantum-photonic information applications^{[8](#page-10-6)}. We perform a proof-of-concept experiment by creation of large photon number states, interfering them on a beam splitter and using photon-counting detection. Existing quantum technologies may scale it up towards diverse applications.

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Signal extraction, compression and analysis in diagnostics, astronomy, chemistry and digital broadcasting often builds on effective implementation of the discrete Fourier transform $(DFT)^2$ $(DFT)^2$. It converts data, a function of e.g. frequency, into their constituent temporal or spatial parts. The DFT is an efficient approximation to the Fourier transform (FT). The signal (x_0, x_1, \ldots, x_S) is taken to be samples of one period of a continuous function, and is turned into a new sequence (X_0, X_1, \ldots, X_S) where

$$
X_k = \frac{1}{\sqrt{S+1}} \sum_{l=0}^{S} e^{-i2\pi \frac{kl}{S+1}} \cdot x_l, \quad k = 0, \dots, S.
$$
 (1)

The DFT does not, however, reproduce all essential features of the FT. In some cases, a transform which is a fractional power of the FT, the α -fractional FT where $0 \le \alpha \le 1$, yields advantages^{[9](#page-10-7)}. For $\alpha = 0$ this transform is the identity, while for $\alpha = 1$ this is the FT. The α fractional DFT defined as the α power of Eq. [\(1\)](#page-1-0) does not correspond to the α -fractional FT^{[9](#page-10-7)}.

The DFT is powerful due to the fast Fourier transform algorithm $(FFT)^2$ $(FFT)^2$. Using an FFT lowers the number of operations from $O(2^{2n})$ to $O(n2^n)$ which nevertheless remains a bottleneck in signal processing^{[10](#page-11-0)}. The FFT employs a "divide and conquer" method to recursively split Eq. (1) into 2^n sums which can be processed quickly, and therefore is applicable to signals of period 2^n . Notably, the minimal number of operations required to implement the DFT is unknown^{[11](#page-11-1)}. The quantum Fourier transform (QFT), the cornerstone of quantum algorithms^{[12,](#page-11-2) [13](#page-11-3)}, enables implementation of the DFT on quantum amplitudes with $O(n \log n)$ operations by processing n qubits $(n \log n)$ quantum bits encode 2^n amplitudes)^{[14](#page-11-4)}.

In many applications, e.g. bioimaging, the signals are typically not periodic and are random in length. For such cases, the Kravchuk transform (KT) is a useful alternative to the FFT because it can be applied to finite signal processing $9,16$ $9,16$. The KT computes orthogonal moments correspond-

ing to the Kravchuk polynomials, which are discrete and orthogonal with respect to a binomial distribution in the data space^{[24](#page-42-0)}. By varying a parameter of the binomial distribution, one is able to set the fractionality α of the KT (SI). This feature allows to explore a specific region of interest of an image. To illustrate the action of a KT, the numerical study in Fig. S1 in the SI demonstrates advantages of the KT over FFT in reconstructing test images.

The KT's computational time is equal to the DFT's runtime^{[21](#page-42-1)} (SI) and implementations with lower number of operations are of high demand. Recently, quantum KTs (QKTs) have been realized in waveguides with two photons, but they are difficult to scale up and their fractionality is fixed by waveguide length^{[5,](#page-10-3)[6](#page-10-4)}.

The α -fractional KT employs the weighted Kravchuk polynomials $\phi_k^{(p)}$ $k^{(p)}(q, S)^{24}$ $k^{(p)}(q, S)^{24}$ $k^{(p)}(q, S)^{24}$ which are real-valued and correspond to wave functions of finite harmonic oscillators

$$
X_k = \sum_{l=0}^{S} e^{-i\frac{\pi\alpha}{2}\frac{S}{2}} e^{i\frac{\pi}{2}(l-k)} \phi_k^{(p)}(l - Sp, S) \cdot x_l, \quad k = 0, \dots, S,
$$
 (2)

where $p = \sin^2\left(\frac{\pi\alpha}{4}\right)$ $\binom{\pi\alpha}{4}$. Unlike plane waves, $e^{-i2\pi\frac{kl}{S+1}}$, the polynomials are defined and orthogonal on a set of $S + 1$ points. This enables one to transform the signal as a finite sequence rather than as an infinite periodic one. In the limit of $S \to \infty$, $\phi_k^{(p)}$ $k^{(p)}(q, S)$ tend to eigenfunctions of quantum harmonic oscillators and the α -fractional KT reproduces the α -fractional FT. Eq. [\(2\)](#page-2-0) can be viewed in terms of overlaps of two spin $S/2$ states, in which they are prepared as eigenstates of S_3 and one undergoes a rotation by angle $\frac{\pi \alpha}{2}$ generated by S_1 , $e^{i\frac{\pi}{2}}$ $\frac{\pi}{2}^{(l-k)}\phi_k^{(p)}$ $_{k}^{(p)}(l - Sp, S) =$ $\langle \frac{S}{2}$ $\frac{S}{2}$; $\frac{S}{2} - k\left|e^{i\frac{\pi\alpha}{2}}\right|$ $\frac{18}{2}S_1\left\vert \frac{S}{2}\right\vert$ $\frac{S}{2}$; $\frac{S}{2} - l$.

In this Letter, we demonstrate a single-step QKT with tunable fractionality using quantum effects, based on multi-particle bosonic interference resulting from an exchange interaction. To this end, we interfere photon number states (light pulses with definite particle number) on a beam

Figure 1: **Photonic implementation of a fractional QKT.** a) HOM interference of photon number states on a variable beam splitter followed by two photon counting detectors, b) Setup: Ti:Sa – titanium-sapphire laser pump (blue), $BS - 50 : 50$ beam splitter, τ – optical phase delay, SPDC – periodically-poled potassium titanyl phosphate (PP-KTP) nonlinear spontaneous parametric down conversion waveguide chip which produces photon number correlated states (red), PBS – polarization beam splitter, VC – variable coupler, TES – transition edge sensors, DAQ – data acquisition unit.

Figure 2: HOM interference and QKT on a Bloch sphere. a-d) Two-mode Fock states (blue) correspond to Dicke states (black) – the basis of spin $\frac{S}{2}$ states. HOM interference turns Dicke states into a superposition of them. This coincides with a rotation $R_{\theta,\varphi}$ in the Dicke state basis. The two most distinct cases are shown: the rotation $R_{\frac{\pi}{2}, \frac{\pi}{2}}$ $\frac{\pi}{2}$ of the pole $\left|\frac{S}{2}\right|$ $\frac{S}{2},\frac{S}{2}$ $\frac{S}{2}$ and of the great circle state $\left|\frac{S}{2}\right|$ $\frac{S}{2}$, 0). e-h) Q-function representation of a-d. HOM interference implements a rotation on the Bloch sphere by $\theta = \frac{\pi}{2}$ $\frac{\pi}{2}$ around S_x of input S_z -eigenbasis Dicke states and thus, the full QKT, cf. Eq. [\(2\)](#page-2-0). The sequence (x_0, x_1, \ldots, x_S) is $(1, 0, 0, \ldots, 0)$ in a) and $(0, \ldots, 1, \ldots, 0)$ in c). The QKT transfers the input – a position eigenstate – into the same state but in S_y basis – a momentum eigenstate.

splitter (BS) with an adjustable splitting ratio. This leads to a multi-particle Hong–Ou–Mandel (HOM) effect^{[18](#page-11-6)} which we observe for states with up to five photons. This QKT implementation enables constant-time quantum information processing for qudit data encoding which is set by the total number of interfering particles S, allowing up to $d = S+1$ signal samples.

Photon number (Fock) states $|l\rangle = \frac{(a^{\dagger})^l}{\sqrt{l!}}$ $\frac{a^{\dagger} \, v^l}{\sqrt{l!}} |0\rangle$ and $|S - l\rangle = \frac{(b^{\dagger})^{S-l}}{\sqrt{(S-l)!}} |0\rangle$ impinging on a beam splitter (BS) exhibit a generalized HOM effect, Fig. [1a](#page-15-0). A BS interaction between two such inputs described by annihilation operators a and b is $U_{BS} = \exp\left\{\frac{\theta}{2}\right\}$ $\frac{\theta}{2}(a^{\dagger}be^{-i\varphi}-ab^{\dagger}e^{i\varphi})\},$ where $r=\sin^2\frac{\theta}{2}$ is the BS reflectivity (defined as the probability of reflection of a single photon) and φ is the phase difference between the reflected and transmitted fields^{[23](#page-42-2)}. Since φ does not influence our experiments, we assume $\varphi = \frac{\pi}{2}$ $\frac{\pi}{2}$ for convenience. If the BS is balanced ($r = 0.5$), two photons at the input ports will leave through the same exit port. This is known as photon bunching^{[7](#page-10-5)}. Similar effects hold for multiphoton number states^{[18](#page-11-6)}. This is reflected in the probability amplitudes of detecting $|k\rangle$ and $|S - k\rangle$ behind the BS, $\mathcal{A}_{S}^{(r)}$ $S^{(r)}(k, l) = e^{-i\theta \frac{S}{2}} \langle k, S - k | U_{BS} | l, S - l \rangle$. This is important for implementing the KT, since $\mathcal{A}_{S}^{(r)}$ $S^{(r)}(k,l) = e^{-i\theta \frac{S}{2}} e^{i\frac{\pi}{2}}$ $\frac{\pi}{2}^{(l-k)} \cdot \phi_k^{(r)}$ $\binom{r}{k}(l - Sr, S)$; thus, if we send a quantum state $|\Psi\rangle = \sum_{l=0}^{S} x_l |l, S - l\rangle$ into the BS, the probability of measuring k and $S - k$ photons behind is the absolute square of a fractional QKT of the input probability amplitudes, $|X_k|^2 = |\sum_{l=0}^S \mathcal{A}_S^{(r)}|$ $S^{(r)}(k, l) \cdot x_l|^2$, cf. Eq. [\(2\)](#page-2-0). The reflectivity r determines the QKT fractionality, $\alpha = \frac{2\theta}{\pi} = \frac{4}{\pi}$ $\frac{4}{\pi}$ arcsin \sqrt{r} . Since two-mode optical interference can be achieved in a single step, regardless of the number of photons involved, this process implements a constant time QKT. For full derivations see SI.

A deeper understanding of the result may be gained from the Schwinger representation of the spin algebra (SI) which links multiphoton interference to spin systems and allows the quantum states to be visualized on a Bloch sphere. In this picture, a total of S photons corresponds to a

spin- $\frac{S}{2}$ system. The Hamiltonian generating $U_{BS} = \exp\{-i\theta H_{BS}\}$ corresponds to an S_x operator for a spin- $\frac{S}{2}$. The two-mode Fock state $|l, S - l\rangle$ corresponds an $S_z = \frac{S}{2} - l$ eigenstate, known as a Dicke state. Hence, HOM interference may be considered a rotation $R_{\theta,\varphi} = \exp\{-i\theta S_x\}$ of S_z around the S_x axis on the sphere. It transfers the eigenstate $\frac{S}{2}$ $\frac{S}{2}$; $\frac{S}{2} - l$ to a superposition of Dicke states, Figs. [2a](#page-32-0)-d. The Q-function in Figs. [2e](#page-32-0)-h shows that the initial and final states are eigenstates of two complementary observables, S_z and S_y , respectively. Thus, one may identify the former with a position, while the latter with a momentum eigenstate.

The experimental setup for multiphoton HOM interference is depicted in Fig. [1b](#page-15-0). Two pulsed spontaneous parametric down-conversion (SPDC) sources each generate two-mode photonnumber correlated states (see Methods). The signal and idler are separated with a polarization BS (PBS) into four spatial modes. The modes A and D are used for heralding and creation of Fock states $|l\rangle$ in B and $|S - l\rangle$ in C which interfere in a variable ratio fiber coupler (the BS). An optical path delay τ in one of the pump beams ensures optimal temporal overlap at interference. Photon-number-resolved measurements are achieved using transition edge sensors (TESs) that we previously estimated to achieve over 90% efficiency^{[29](#page-43-0)}.

We interfered the vacuum $|0\rangle$ (l = 0) with multiphoton Fock states $|S\rangle$ (S – l = S) on a coupler with splitting ratios $r = 0.05$ (green), 0.2 (red), 0.5 (blue) and 0.95 (gray), and measured photon number statistics. They are depicted in Figs. [3a](#page-37-0)-c for $S = 3, 4, 5$. The input states encode sequences $(x_0 = 1, x_1 = 0, \ldots, x_s = 0)$, while the measured probabilities set their QKTs: $(|X_0|^2, |X_1|^2, \ldots, |X_S|^2)$, where $|X_k|^2 = |\mathcal{A}_S^{(r)}(k, 0)|^2$. The reflectivities used correspond to fractionalities $\alpha = 0.28, 0.60, 1.00, 1.72$. Errors were estimated as a square root inverse of the number of measurements (SI). The second-order interferometric visibility reached values between 71.4% and 98.6% for $S = 5$ (SI).

Figure 3: Photon number statistics resulting from Fock state $|l, S - l\rangle$ interference. The probabilities of detecting $|k\rangle$ and $|S - k\rangle$ photons behind the BS for input a) $|0, 3\rangle$, b) $|0, 4\rangle$, c) $|0, 5\rangle$, d) $|1, 2\rangle$, e) $|2, 2\rangle$, f) $|2, 3\rangle$. The BS reflectivities are $r = 0.05$ (green), 0.2 (red), 0.5 (blue) and 0.95 (gray). Vertical bars represent theoretical values for an ideal system, while dots are values determined in experiment. The states a)-c) encode sequences $(x_0 = 1, x_1 = 0, \ldots, x_S = 0)$, and in d) – $(0, 1, 0, 0)$, c) – $(0, 0, 1, 0, 0)$, f) – $(0, 0, 1, 0, 0, 0)$, respectively. The measured probabilities set their QKTs $(|X_0|^2, |X_1|^2, \dots, |X_S|^2), |X_k|^2 = |\sum_{l=0}^{S} A_S^{(r)}|$ $\int_{S}^{(r)}(k,l)\cdot x_l|^2$ of fractionality $\alpha = 0.28$ (green), 0.60 (red), 1.00 (blue) and 1.72 (gray).

For the same values of r we measured photon number distribution resulting from interference of $|1, 2\rangle$, $|2, 2\rangle$ and $|2, 3\rangle$. They are shown in Figs. [3d](#page-37-0)-f. The inputs encode $(0, 1, 0, 0, 0, 0, 1, 0, 0)$, $(0, 0, 1, 0, 0, 0)$, while $|X_k|^2 = |\mathcal{A}_3^{(r)}(k, 1)|^2$, $|\mathcal{A}_4^{(r)}(k, 2)|^2$ and $|\mathcal{A}_5^{(r)}(k, 2)|^2$, respectively. The visibility was between 54.8% and 99.5% ($S = 5$) (SI).

Fig. [3](#page-37-0) shows that the theoretical values computed for an ideal system (the bars) match the experimental results (the dots) well.

Realization of the fractional QKT with qudit systems opens a new prospect for transformation of large data sequences in $O(1)$ time. This is not possible with the implementations based on waveguides. Both cases are examples of a non-universal quantum computer optimized for one task which is the basis for a variety of important applications^{[9](#page-40-0)}. The photonic proof of concept is currently limited by the range of input states that can be prepared. However, deterministic creation of an arbitrary superposition of Fock states has been demonstrated for trapped ions and supercon-ducting resonators^{[21](#page-12-0)}. Since a BS sees orthogonal spectral or polarization modes independently, one can extend the transform to higher dimensions $^{22, 23}$ $^{22, 23}$ $^{22, 23}$ $^{22, 23}$ $^{22, 23}$. We note that the QKT could also be im-plemented on existing quantum annealing processors^{[24](#page-12-3)}, which operate on a chain of interacting spin- $\frac{1}{2}$ systems (SI), and using HOM interference of fermions with symmetric wavefunction of the interfering degrees of freedom.

Our result, along with the fact that qudit-based algorithms exhibit significantly lower num-ber of operations than qubit-based ones^{[25](#page-12-4)}, motivates the further development of highly-controllable quantum harmonic oscillator platforms with implications for quantum signal processing in a whole range of applications. Provided efficient input state preparation and detection of larger Fock states, the $O(1)$ QKT demonstrated here in principle may find practical applications in imaging of unprecedented quality, fostering early diagnostics and neuroscience^{[5](#page-40-1)}.

Methods

A light pulse from a Ti:Sapphire laser at 775 nm (FWHM of 2 nm; repetition rate of 75 kHz) pumps collinear type-II phase-matched 8 mm-long SPDC waveguides written in a periodically poled KTP (PP-KTP) crystal sample. They generate two independent photon-number correlated states – the two-mode squeezed vacua $|\Psi\rangle = \sum_{n=0}^{\infty} \lambda_n |n,n\rangle$, where $\lambda_n = \frac{\tanh^n g}{\cosh g}$ $\frac{\text{ann'' } g}{\cosh g}$ is a probability amplitude for creation of a pair of n photons and g is the parametric gain. The average photon number in the signal and idler mode equals $\langle \hat{n} \rangle = \sinh^2 g$. For small g, $\cosh g \approx 1$, and thus $\lambda_n^2 \approx \sinh^{2n} g = \langle \hat{n} \rangle^n$. In the experiment, the average photon number is $\langle \hat{n} \rangle \approx 0.2$. This value is sufficient to ensure the emission of multiphoton pairs, but at the same time to diminish the interferometric visibility of two-photon events. In both output states, the signal and idler pulses are split with a polarization beam-splitter (PBS) to four spatial modes $A-D$. Subsequently, they are filtered by bandpass filters with 3 nm FWHM angle-tuned to the central wavelength of their respective spectra, in order to reduce the broadband background typically generated in dielectric nonlinear waveguides^{[27](#page-12-5)}. The pump beam is discarded with an edge filter. The modes A and D are used for heralding and conditional creation of Fock states in modes B and C which interfere in a variable ratio PM fiber coupler. The coupling ratio can be set in the range 0-100% with an error of $\pm 1.5\%$. The heralding signal modes (H-pol.) are centered at 1554 nm, while the interfering idler modes (V-pol.) are at 1546 nm. We employ transition-edge sensors (TES) running at 70 mK which allow for photon-number resolved measurements in all modes^{[31](#page-43-1)}. Their voltage output is captured with an ADC card.

Before demonstrating the HOM interference, we characterized the setup. A high photon number resolution and single-mode input states are pivotal for this experiment. The resolution

of TES detectors (the confidence that the detector gives a correct information about the number of photons) was previously confirmed to exceed $95\%^{29}$ $95\%^{29}$ $95\%^{29}$. The depth of the HOM dip of $85.9\% \pm$ 0.3% for a two-photon interference indicates an effective Schmidt mode number $K = 1.16$ (SI). For the measured 4-tuples of photon numbers losses were computed by assuming perfect setup components, each followed by a beam splitter with a reflection coefficient introducing the loss. We estimate the total transmission in each mode to be approximately 50%. For the details, see SI.

Measurements for individual settings of the splitting ratio were taken over approximately 400 seconds, giving 10^9 data samples for each r ranging from 0 to 1 with a step of approximately 3%. Small error bars for low photon numbers and larger bars for the higher ones result from keeping the pump power fixed and near-single-modeness of the interfering beams.

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Supplemental Information

1 Applications of the Kravchuk transform

Comparison of the KT and FFT for reconstruction of MRI scans. Medical imaging, especially Magnetic Resonance Imaging (MRI) is an example of the most prominent applications of the FFT algorithm^{[1](#page-40-2)}. The signal received from electromagnetic coils is encoded in the frequency domain (k-space) and must be processed by the inverse Fourier transform in order to obtain a readable scan of the patient's body^{[2](#page-40-3)}. The MRI resolution and thus, the quality of the imaging is strongly limited by the bandwidth of the receiver and the accuracy of the magnetic field^{[3](#page-40-4)}. The signal is sampled with an analog-to-digital converter (ADC) and stored in the computer memory before processing. Typical sampling capacity of the ADC corresponds to approximately 250 frequency slots and the magnetic gradient is steered with an accuracy of 100-150 levels, corresponding to the effective size of voxels to be several millimeters^{[3](#page-40-4)}. Further improvement of the accuracy is very difficult and requires higher magnetic fields as well as longer examination time which are not indifferent to the patient's health 4 . This is the reason why the accuracy of data processing is crucial. It is worth noting that the voxel size must be of the order of micrometers to examine neurons^{[5](#page-40-1)}.

Fig. [S1](#page-15-0) shows a comparative numerical study of the KT and FFT for a "pirate" test image and a brain scan from the OASIS database^{[6](#page-40-6)}. It highlights the influence of sample size and added noise on the reconstruction quality. The FFT produces artifacts due to zero padding, while some details (which could be tumor cells) are missing. This is best captured by the structural similarity index $(SSIM)^7$ $(SSIM)^7$. For the "pirate" it was 0.74 (FFT) and 0.96 (KT), and for the brain it was 0.81 and 0.97, respectively. The mean square error $(MSE)^7$ $(MSE)^7$ was ten times smaller and the peak signal-to-noise ratio (PSNR)^{[7](#page-40-7)} was 10 dB larger for the KT than for the FFT. The FFT leads to degradation of the usable resolution from 1-2 mm per voxel to over 5 mm. Our findings confirm the results of the previous research^{[8](#page-40-8)}.

Figure S1: **Numerical example of FFT and KT image processing**. Two 257×257 -pixel test images, a "pirate" and a brain scan from the OASIS database, were converted to the frequency domain with the KT and FFT (full image moment orders), supplemented with a 1% additive white Gaussian noise, and then reconstructed back with corresponding inverse transforms (without any truncation) to model the operation of an MRI analysis: a) & d) the source data, b) & e) the images processed with the KT. The green circles mark some fine details which were retained during this processing, c) & f) the images processed with the FFT after zero-padding to $2^n \times 2^n = 512 \times 512$ pixels. The red circles highlight the artifacts resulting from zero padding.

Digital image processing and computer vision. The image analysis with Kravchuk moments was first proposed by Yap et al.^{[9](#page-40-0)}, where the transform coefficients were used as data vectors for shape recognition. The new method overperformed the "classical" Hu approach^{[10](#page-40-9)} giving accuracy over 80% for noisy input, compared to 31% obtained with the Hu's method. The Kravchuk moments were thoroughly compared with the state-of-the-art methods towards various medical imaging applications^{[8](#page-40-8)}. Tests were performed with magnetic resonance data coming from open repositories of brain and knee examinations. The images were reconstructed with the Kravchuk, Zernike, Pseudo-Zernike, Fourier-Merlin, Legendre and Chebyshev kernels^{[8,](#page-40-8)[9](#page-40-0)}. It has been pointed out that only the Kravchuk and Chebyshev transforms are discrete and allow to operate in the original Cartesian image coordinates $8,11$ $8,11$. The Kravchuk-based method presented the best behavior in most test cases, giving the smallest reconstruction error and the highest peak signal-to-noise ratio as the moment order increased thus, is best suited for processing of high resolution data $8,11$ $8,11$. For similar reasons, the numerical method of the Kravchuk moments was chosen as the most promis-ing in analysis of breast mammography images^{[12](#page-41-1)}, where it allowed to identify benign and malign masses with 90% accuracy, compared to 81% offered by the other techniques. This scheme was proven to outperform also other algorithms in analysis of computer tomography and ultrasound scans towards recognition of liver and prostate tumors^{[13](#page-41-2)}.

The Kravchuk transform is also getting attention in processing of other kinds of numerical data, for example in the Chinese character recognition^{[14](#page-41-3)}, development of error-correcting codes^{[15](#page-41-4)} and watermarking in anti-fraud techniques^{[16](#page-41-5)} (the fractional KT).

KT software implementations. The algorithms for computation of the Kravchuk polynomials and transforms have been proposed both for software^{[17](#page-41-6)} and hardware solutions^{[18](#page-41-7)}. They are in-cluded in various programming libraries^{[19](#page-42-3)}, e.g. POLPAK, an established package of polynomial-computing routines^{[20](#page-42-4)}. However, these algorithms are still underperforming the FFT because their

number of operations is $O(n^2 \log^2 n)$ compared to $O(n \log n)$ for the FFT. This turns into the re-construction time of e.g. 15 min per image instead of several seconds^{[19](#page-42-3)}. This seriously limits the scope of the KT in the numerical image processing. This result has been improved by Venkataramana et al. to $O(n^2)$ by using the Clenshaw's recurrence formula^{[21](#page-42-1)}.

2 Theory: multi-photon Hong–Ou–Mandel interference

We will now analyze in a detailed manner a generalized multi-photon HOM effect. As explained in the main text, we consider two interfering modes a and b on a beam splitter device with a tunable reflectivity r (defined as the probability of reflection of a single photon). As the input states we take photon number (Fock) states $|l\rangle_a = \frac{(a^{\dagger})^l}{\sqrt{l!}}$ $\frac{a^{\dagger}\right)^l}{\sqrt{l!}}\ket{0}_a$ and $\ket{S-l}_b = \frac{(b^{\dagger})^{S-l}}{\sqrt{(S-l)!}}\ket{0}_b$.

The Schwinger representation. One may represent the su(2) Lie algebra in terms of the annihila-tion and creation operators of the harmonic oscillator – the Schwinger representation^{[22](#page-42-5)}. For a single spin two independent oscillators a and b are required. The spin operators are then constructed in the following way

$$
S_x = \frac{a^{\dagger}b + a b^{\dagger}}{2}, \qquad S_y = \frac{i(a b^{\dagger} - a^{\dagger}b)}{2}, \qquad S_z = \frac{a^{\dagger}a - b^{\dagger}b}{2}, \qquad S_0 = \frac{a^{\dagger}a + b^{\dagger}b}{2}.
$$
 (S1)

 S_0 is the Casimir operator $S_0(S_0 + 1) = S_x^2 + S_y^2 + S_z^2$. The spin components fulfill the standard su(2) commutation relations

$$
[S_x, S_y] = iS_z, \t [S_y, S_z] = iS_x, \t [S_z, S_x] = iS_y.
$$
 (S2)

Beam splitter. Interference of two independent modes a and b on a beam splitter is governed by the following Hamiltonian

$$
H = H_0 + H_{BS},\tag{S3}
$$

$$
H_0 = \frac{\hbar}{2} \left(a^\dagger a + b^\dagger b \right),\tag{S4}
$$

$$
H_{BS} = \frac{i\hbar}{2} \left(a^{\dagger} b \, e^{-i\varphi} - a \, b^{\dagger} e^{i\varphi} \right). \tag{S5}
$$

 H_0 is the free quantum oscillator energy and H_{BS} – the beam splitter interaction^{[23](#page-42-2)}. φ is the phase difference between the reflected and transmitted fields behind the beam splitter. H_0 commutes with H_{BS} .

Using the Schwinger representation, we express H in terms of the spin operators S_0 , S_x , S_y , S_z

$$
H_0 = \hbar S_0,\tag{S6}
$$

$$
a^{\dagger}b e^{-i\varphi} - a b^{\dagger} e^{i\varphi} = \cos\varphi \left(a^{\dagger} b - a b^{\dagger} \right) - i \sin\varphi \left(a^{\dagger} b + a b^{\dagger} \right)
$$
 (S7)

$$
= 2i\left(\cos\varphi \cdot S_y - \sin\varphi \cdot S_x\right). \tag{S8}
$$

$$
H_{BS} = \frac{i\hbar}{2} 2i \left(\cos \varphi \cdot S_y - \sin \varphi \cdot S_x \right) \tag{S9}
$$

$$
= \hbar \left(\sin \varphi \cdot S_x - \cos \varphi \cdot S_y \right). \tag{S10}
$$

The Hamiltonian generates the evolution operator

$$
U = \exp\{-i\theta H/\hbar\} \tag{S11}
$$

$$
= \exp\{-i\theta(H_0 + H_{BS})/\hbar\} \tag{S12}
$$

$$
= \exp\{-i\theta \, H_{BS}/\hbar\} \exp\{-i\theta \, H_0/\hbar\} \tag{S13}
$$

$$
=U_{BS}U_0,\t\t(S14)
$$

$$
U_0 = \exp\left\{-i\theta S_0\right\},\tag{S15}
$$

$$
U_{BS} = \exp\left\{-i\theta\left(\sin\varphi \cdot S_x - \cos\varphi \cdot S_y\right)\right\}.
$$
 (S16)

The evolution in the Heisenberg picture allows to establish a linear relation between the input (a, b) and the output (a_r, a_t) annihilation operators

$$
a_r = U_{BS}^{\dagger} \, a \, U_{BS} = a \cos \frac{\theta}{2} + b \, e^{-i\varphi} \sin \frac{\theta}{2},\tag{S17}
$$

$$
a_t = U_{BS}^{\dagger} b U_{BS} = -a e^{i\varphi} \sin\frac{\theta}{2} + b \cos\frac{\theta}{2}.
$$
 (S18)

The relation takes the following matrix form

$$
\mathbf{U}_{\mathbf{B}\mathbf{S}} = \begin{pmatrix} \cos\frac{\theta}{2} & \sin\frac{\theta}{2}e^{-i\varphi} \\ -\sin\frac{\theta}{2}e^{i\varphi} & \cos\frac{\theta}{2} \end{pmatrix},\tag{S19}
$$

where $U_{BS}U_{BS}^{\dagger}=1$ and $U_{BS}^{\dagger}=U_{BS}^{-1}$ hold true. U_0 amounts to a global phase.

We now substitute $\sin \frac{\theta}{2} = \sqrt{r}$ and $\cos \frac{\theta}{2} = \sqrt{1-r}$ to relate the evolution directly to the beam splitter reflectivity

$$
\mathbf{U}_{\mathbf{B}\mathbf{S}} = \begin{pmatrix} \sqrt{1-r} & e^{-i\varphi}\sqrt{r} \\ -e^{i\varphi}\sqrt{r} & \sqrt{1-r} \end{pmatrix},
$$
\n(S20)

$$
\mathbf{U}_{\mathbf{B}\mathbf{S}}^{-1} = \mathbf{U}_{\mathbf{B}\mathbf{S}}^{\dagger} = \begin{pmatrix} \sqrt{1-r} & -e^{-i\varphi}\sqrt{r} \\ e^{i\varphi}\sqrt{r} & \sqrt{1-r} \end{pmatrix} . \tag{S21}
$$

This brings us to the following relation between the input and output creation operators, to be used in the next section

$$
\begin{pmatrix} a \\ b \end{pmatrix} = \mathbf{U}_{\mathbf{B}\mathbf{S}}^{-1} \begin{pmatrix} a_r \\ a_t \end{pmatrix} = \begin{pmatrix} \sqrt{1-r} & -e^{-i\varphi}\sqrt{r} \\ e^{i\varphi}\sqrt{r} & \sqrt{1-r} \end{pmatrix} \begin{pmatrix} a_r \\ a_t \end{pmatrix},
$$
(S22)

$$
\begin{pmatrix} a^{\dagger} \\ b^{\dagger} \end{pmatrix} = \begin{pmatrix} \sqrt{1-r} & -e^{i\varphi}\sqrt{r} \\ e^{-i\varphi}\sqrt{r} & \sqrt{1-r} \end{pmatrix} \begin{pmatrix} a_r^{\dagger} \\ a_t^{\dagger} \end{pmatrix},
$$
(S23)

$$
a^{\dagger} = \sqrt{1 - r} a_r^{\dagger} - e^{i\varphi} \sqrt{r} a_t^{\dagger}, \qquad (S24)
$$

$$
b^{\dagger} = e^{-i\varphi} \sqrt{r} a_r^{\dagger} + \sqrt{1 - r} a_t^{\dagger}.
$$
 (S25)

Photon number amplitude. Let the input states in modes a and b be the Fock states $|l\rangle$ and $|S - l\rangle$, respectively. Then,

$$
U_0 |l\rangle_a |S - l\rangle_b = e^{-i\theta \frac{S}{2}} |l\rangle_a |S - l\rangle_b, \qquad (S26)
$$

$$
U_{BS} |l\rangle_a |S - l\rangle_b = U_{BS} \frac{\left(a^{\dagger}\right)^l}{\sqrt{l!}} \frac{\left(b^{\dagger}\right)^{S-l}}{\sqrt{(S-l)!}} |0\rangle
$$
\n
$$
(S27)
$$

$$
=\frac{1}{\sqrt{l!(S-l)!}}\left(\sqrt{1-r}a_r^{\dagger}-e^{i\varphi}\sqrt{r}a_t^{\dagger}\right)^l\left(e^{-i\varphi}\sqrt{r}a_r^{\dagger}+\sqrt{1-r}a_t^{\dagger}\right)^{S-l}|0\rangle
$$
\n(S28)

$$
= \frac{1}{\sqrt{l!(S-l)!}} \sum_{m=0}^{l} \sum_{n=0}^{S-l} {l \choose m} {S-l \choose n} \left(\sqrt{1-r} a_r^{\dagger}\right)^m \left(-e^{i\varphi} \sqrt{r} a_t^{\dagger}\right)^{l-m} \times
$$

$$
\times \left(e^{-i\varphi} \sqrt{r} a_r^{\dagger}\right)^n \left(\sqrt{1-r} a_t^{\dagger}\right)^{S-l-n} |0\rangle
$$
(S29)

$$
= \frac{1}{\sqrt{l!(S-l)!}} \sum_{m=0}^{l} \sum_{n=0}^{S-l} {l \choose m} {S-l \choose n} \left(-e^{i\varphi} \sqrt{r} \right)^l \left(\sqrt{1-r} \right)^{S-l} \times \qquad (S30)
$$

$$
\times \left(\sqrt{1-r} \right)^{m-n} \left(\sqrt{r} \right)^{n-m} \left(e^{-i\varphi} \right)^{m+n} \times \qquad \qquad \times (-1)^{-m} \left(a_r^{\dagger} \right)^{m+n} \left(a_r^{\dagger} \right)^{S-m-n} |0\rangle
$$

$$
U |l\rangle_a |S-l\rangle_b = e^{-i\theta \frac{S}{2}} \frac{\left(-e^{i\varphi} \sqrt{r} \right)^l \left(\sqrt{1-r} \right)^{S-l}}{\sqrt{l!(S-l)!}} \sum_{m=0}^{l} \sum_{n=0}^{S-l} {l \choose m} {S-l \choose n} (-1)^{-m} \times \qquad \qquad \times (e^{-i\varphi})^{m+n} \left(\sqrt{\frac{1-r}{r}} \right)^{m-n} \times \qquad \qquad \times \sqrt{(m+n)!(S-m-n)!} \left| m+n, S-m-n \right\rangle . \qquad (S31)
$$

Let us substitute $m + n = k$ to change the summation variables. Then,

$$
|m+n, S-m-n\rangle = |k, S-k\rangle
$$

and the ranges of k and m are as follows

$$
0 \le m + n = k \le S
$$

\n
$$
0 \le k - m = n \le S - l \Rightarrow k + l - S \le m \le k,
$$

\n
$$
\sum_{m=0}^{l} \sum_{n=0}^{S-l} \Rightarrow \sum_{k=0}^{S} \sum_{m=\max\{0,k+l-S\}}^{\min\{l,k\}},
$$

\n
$$
U|l\rangle_a|S-l\rangle_b = e^{-i\theta \frac{S}{2}} \frac{(-e^{i\varphi}\sqrt{r})^l(\sqrt{1-r})^{S-l}}{\sqrt{l!(S-l)!}} \sum_{k=0}^{S} \sum_{m=\max\{0,k+l-S\}}^{\min\{l,k\}} {l \choose m} {S-l \choose k-m} (-1)^{-m} \times
$$

\n
$$
\times (e^{-i\varphi})^k \left(\sqrt{\frac{1-r}{r}}\right)^{2m-k} \sqrt{k!(S-k)!} |k, S-k\rangle.
$$

\n(S32)

The probability amplitude of detecting k and $S-k$ photons behind the beam splitter provided that l and $S - l$ were injected into it is

$$
\mathcal{A}_S(k,l) = \langle k, S-k | U | l, S-l \rangle, \qquad (S33)
$$

thus,

$$
U|l\rangle_a|S-l\rangle_b = \sum_{k=0}^{S} \mathcal{A}_S(k,l)|k,S-k\rangle, \qquad (S34)
$$

where

$$
\mathcal{A}_{S}(k,l) = e^{-i\theta} \frac{S}{2} \frac{\left(-e^{i\varphi}\sqrt{r}\right)^{l}\left(\sqrt{1-r}\right)^{S-l}}{\sqrt{l!(S-l)!}} \sum_{m=\max\{0,k+l-S\}}^{m\text{in}\{l,k\}} \binom{l}{m} \binom{S-l}{k-m} (-1)^{-m} \left(e^{-i\varphi}\right)^{k} \times \left(\sqrt{\frac{1-r}{r}}\right)^{2m-k} \sqrt{k!(S-k)!}
$$
\n
$$
\times \left(\sqrt{\frac{1-r}{r}}\right)^{2m-k} \sqrt{k!(S-k)!}
$$
\n(S35)

$$
= e^{-i\theta} \frac{S}{2} \frac{\left(-e^{i\varphi}\sqrt{r}\right)^l \left(\sqrt{1-r}\right)^{S-l}}{\sqrt{l!(S-l)!}} \left(e^{-i\varphi}\right)^k \left(\sqrt{\frac{1-r}{r}}\right)^{-k} \sqrt{k!\left(S-k\right)!} \times \sum_{m=\max\{0,k+l-S\}}^{\min\{l,k\}} \binom{l}{m} \binom{S-l}{k-m} (-1)^{-m} \left(\frac{1-r}{r}\right)^m.
$$
\n
$$
(S36)
$$

The inner sum over m in Eq. [\(S36\)](#page-22-0) is a hypergeometric series. In order to simplify it, the identities from Section [3](#page-29-0) are used. The four cases below (A-D) correspond to different summation ranges. For simplicity, let us assume that $l \leq S - l$, i.e. $l \leq \frac{S}{2}$ $\frac{S}{2}$.

Case A: min $\{l, k\} = l$ and max $\{0, k + l - S\} = 0$. This implies $l \le k \le S - l$.

$$
\sum_{m=0}^{l} \binom{l}{m} \binom{S-l}{k-m} (-1)^{-m} \left(\frac{1-r}{r}\right)^m \tag{S37}
$$

$$
= \sum_{m=0}^{l} {l \choose m} {S-l \choose k-m} \left(1 - \frac{1}{r}\right)^m
$$
\n
$$
(S38)
$$

$$
= \sum_{m=0}^{l} {l \choose m} \frac{(S-l)!}{(k-m)!(S-l-k+m)!} \frac{(S-l-k)!k!}{(S-l-k)!k!} (1 - \frac{1}{r})^m
$$
(S39)

$$
= \sum_{m=0}^{l} {l \choose m} \frac{(S-l)!}{(S-l-k)! \, k!} \frac{(S-l-k)!}{(S-l-k+m)!} \frac{k!}{(k-m)!} \left(1 - \frac{1}{r}\right)^m
$$
(S40)

$$
= \sum_{m=0}^{l} {l \choose m} {S-l \choose k} \frac{(-1)^m (-k)_k}{(S-l-k+1)_k} \left(1 - \frac{1}{r}\right)^m \qquad \text{cf. (S92), (S93) (S41)}
$$

$$
= {S-l \choose k} {}_{2}F_{1} [-l, -k; S-l-k+1; 1-\frac{1}{r}] \qquad \qquad \text{cf. (S95)} \qquad (S42)
$$

$$
= \binom{S-l}{k} \frac{(S-l-k+1+k)_K}{(S-l-k+1)_K} {}_2F_1 \left[-l, -k; -S; \frac{1}{r} \right] \qquad \text{cf. (S96)} \quad (S43)
$$

$$
= \binom{S}{k} {}_2F_1 \left[-l, -k; -S; \frac{1}{r} \right]. \tag{S44}
$$

Case B: min $\{l, k\} = k$ and max $\{0, k + l - S\} = k + l - S$. This implies $S - l \le k \le l$, i.e. the empty set.

Case C: min{l, k} = k and max{0, k + l - S} = 0. This implies $k \le l \le S - l$.

$$
\sum_{m=0}^{k} \binom{l}{m} \binom{S-l}{k-m} (-1)^{-m} \left(\frac{1-r}{r}\right)^m \tag{S45}
$$

$$
= \sum_{m=0}^{k} {l \choose m} {S-l \choose k-m} (1 - \frac{1}{r})^m
$$
\n
$$
(S46)
$$

$$
= \sum_{m=0}^{k} \frac{l!}{m! (l-m)!} \frac{(S-l)!}{(k-m)! (S-l-k+m)!} \frac{k! (S-l-k)!}{k! (S-l-k)!} (1 - \frac{1}{r})^m
$$
(S47)

$$
= \sum_{m=0}^{k} \frac{k!}{m! (k-m)!} \frac{(S-l)!}{k! (S-l-k)!} \frac{l!}{(l-m)!} \frac{(S-l-k)!}{(S-l-k+m)!} \left(1 - \frac{1}{r}\right)^m
$$
(S48)

$$
= \sum_{m=0}^{k} {k \choose m} {S-l \choose k} \frac{(-1)^m (-l)_k}{(S-l-k+1)_k} \left(1 - \frac{1}{r}\right)^m
$$
 cf. (S92), (S93)

(S49)

$$
= {S-l \choose k} {}_2F_1 [-k, -l; S-l-k+1; 1-\tfrac{1}{r}]
$$
 cf. (S95)

$$
(S50)
$$

$$
= {S-l \choose k} {}_2F_1 [-l, -k; S-l-k+1; 1-\tfrac{1}{r}]
$$
 cf. (S94)

(S51)

$$
= {S-l \choose k} \frac{(S-l-k+1+k)_K}{(S-l-k+1)_K} {}_2F_1 [-l, -k; -S; \frac{1}{r}]
$$
 cf. (S96)

(S52)

$$
= {S \choose k} {}_2F_1 \left[-l, -k; -S; \frac{1}{r} \right]
$$
\n
$$
(S53)
$$

Case D: min $\{l, k\} = l$ and max $\{0, k + l - S\} = k + l - S$. This implies $l \leq S - l \leq k$. To compute the sum, the following substitution is used: $m = l - m'$.

$$
\sum_{m=k+l-S}^{l} {l \choose m} {S-l \choose k-m} (1-\tfrac{1}{r})^m = \sum_{m'=0}^{S-k} {l \choose l-m'} {S-l \choose k+m'-l} (1-\tfrac{1}{r})^{l-m'} \tag{S54}
$$

$$
= (1 - \frac{1}{r})^l \sum_{m'=0}^{S-k} {S-k \choose m'} \frac{m'! (S-k-m')!}{(S-k)!} \frac{l!}{(l-m')!m'!} \times \frac{(S-1)!}{(S-k-m')! (k+m'-l)!} (-1)^{-m'} (\frac{1}{r} - 1)^{-m'}
$$
(S55)

$$
= (1 - \frac{1}{r})^l \sum_{m'=0}^{S-k} {S-k \choose m'} (-1)^{m'} \frac{l!}{(l-m')!} \frac{(S-l)!}{(S-k)!(k+m'-l)!} \left(\frac{1}{r} - 1\right)^{-m'}
$$
(S56)

$$
= \left(1 - \frac{1}{r}\right)^l \sum_{m'=0}^{S-k} {S-k \choose m'} (-1)^{m'} (-1)^{m'} (-l)_{m'} \times \left((S57) \right)
$$
\n
$$
(S-l)! \qquad (k-l)! \qquad m'
$$

$$
\times \frac{(S-l)!}{(S-k)!(k-l+m')!} \frac{(k-l)!}{(k-l)!} \left(\frac{r}{1-r}\right)^{m'}
$$
\n
$$
= \left(1 - \frac{1}{r}\right)^l \sum_{m'=0}^{S-k} {S-k \choose m'} (-1)^{m'} (-l)_{m'} \frac{(S-l)!}{(S-k)!(S-l-S+k)!} \times \text{cf. (S92), (S93)}
$$
\n
$$
\times \frac{(k-l)!}{(k-l+m')!} \left(\frac{r}{r-1}\right)^{m'}
$$

(S58)

$$
= (1 - \frac{1}{r})^{l} {S - l \choose S - k} \sum_{m'=0}^{S-k} {S - k \choose m'} (-1)^{m'} \frac{(-l)_{m'}}{(k - l + 1)_{m'}} \left(\frac{r}{r-1}\right)^{m'}
$$
 cf. (S95)

(S59)

$$
= (1 - \frac{1}{r})^{l} {S - l \choose S - k} {}_{2}F_{1} [-(S - k), -l; k - l + 1; \frac{r}{r-1}]
$$
 cf. (S96)

(S60)

$$
= (1 - \frac{1}{r})^{l} \binom{S - l}{S - k} \frac{(k - l + 1 + l)_{S - k}}{(k - l + 1)_{S - k}} \times \frac{(S61)}{\times {}_{2}F_{1}} \left[-(S - k), -l; 1 - S + k - l - (k - l + 1); \frac{-1}{r - 1} \right]
$$

$$
= (1 - \frac{1}{r})^{l} {S \choose S-k} {}_{2}F_{1} [-(S-k), -l; -S; \frac{-1}{r-1}] \qquad \qquad \text{cf. (S97)}
$$

(S62)

$$
= \underbrace{\left(1 - \frac{1}{r}\right)^l \left(1 - \frac{-1}{r-1}\right)^l}_{=1} {S \choose k} {}_2F_1 \left[-S + (S - k), -l; -S; \frac{\frac{-1}{r-1}}{\frac{-1}{r-1} - 1}\right]
$$
\n(S63)

$$
= \binom{S}{k} {}_2F_1\left[-k, -l; -S; \frac{1}{r}\right]
$$
 cf. (S94)

$$
(S64)
$$

$$
= \binom{S}{k} {}_2F_1 \left[-l, -k; -S; \frac{1}{r} \right]. \tag{S65}
$$

Summarizing, the inner sum in Eq. [\(S36\)](#page-22-0) equals $\binom{S}{k}$ $_{k}^{S}$ ₂ F_{1} $\left[-l,-k;-S;\frac{1}{r}\right]$ $\frac{1}{r}$ under the assumption that $l \leq \frac{S}{2}$ $\frac{S}{2}$. The probability amplitude can be rewritten into the following form

$$
\mathcal{A}_{S}(k,l) = \frac{\left(-e^{i\varphi}\sqrt{r}\right)^{l}\left(\sqrt{1-r}\right)^{S-l}}{\sqrt{l!(S-l)!}}e^{-i\theta\frac{S}{2}}\left(e^{-i\varphi}\right)^{k}\left(\sqrt{\frac{1-r}{r}}\right)^{-k}\sqrt{k!\left(S-k\right)!}\binom{S}{k}{}_{2}F_{1}\left[-l,-k;-S;\frac{1}{r}\right],\tag{S66}
$$

$$
\sqrt{\frac{k!\,(S-k)!}{l!\,(S-l)!}}\binom{S}{k} = \sqrt{\binom{S}{k}\binom{S}{l}},
$$

 $\overline{}$

$$
\mathcal{A}_{S}(k,l) = \sqrt{\binom{S}{k}\binom{S}{l}} (-1)^{l} \left(e^{i\varphi}\right)^{l-k} e^{-i\theta \frac{S}{2}} \left(\sqrt{1-r}\right)^{S} \left(\sqrt{\frac{r}{1-r}}\right)^{l+k} {}_{2}F_{1}\left[-l,-k;-S;\frac{1}{r}\right]
$$
\n(S67)

$$
= \sqrt{\binom{S}{k}\binom{S}{l}} (-1)^l \left(e^{i\varphi}\right)^{l-k} e^{-i\theta \frac{S}{2}} \left(\cos\frac{\theta}{2}\right)^S \left(\tan\frac{\theta}{2}\right)^{l+k} {}_2F_1 \left[-l, -k; -S; \left(\sin\frac{\theta}{2}\right)^{-2}\right]
$$
\n(S68)

$$
= \sqrt{\binom{S}{k}\binom{S}{l}} (-1)^l \left(e^{i\varphi}\right)^{l-k} e^{-i\theta \frac{S}{2}} \left(\cos\frac{\theta}{2}\right)^S \left(\tan\frac{\theta}{2}\right)^{l+k} {}_2F_1 \left[-k, -l; -S; \left(\sin\frac{\theta}{2}\right)^{-2}\right].
$$
\n(S69)

The photon number statistics behind the beam splitter is given by the probability $p_S(k, l)$ = $|\mathcal{A}_S(k,l)|^2$

$$
p_S(k,l) = {S \choose k} {S \choose l} \left(\cos \frac{\theta}{2}\right)^{2S} \left(\tan \frac{\theta}{2}\right)^{2(l+k)} \left|{}_2F_1 \left[-l, -k; -S; \left(\sin \frac{\theta}{2}\right)^{-2}\right]\right|^2 = p_S(l,k). \tag{S70}
$$

Kravchuk transform. The α -fractional Kravchuk transform of an input sequence $x_n = f(\xi_n)$, where $n = 0, 1, ..., N$ and $\xi_n = (n - N/2)$, is defined as follows^{[24](#page-42-0)} (cf. Eq. (5.2))

$$
\mathbf{X}_n = \sum_{n'=0}^{N} F_{n,n'}^{\alpha} x_{n'},
$$
\n(S71)

$$
F_{n,n'}^{\alpha} = e^{i\frac{\pi}{2}(n+n'-N\alpha/2)} \sqrt{\binom{N}{n}\binom{N}{n'}} \cos^N\left(\frac{\pi\alpha}{4}\right) \tan^{n+n'}\left(\frac{\pi\alpha}{4}\right) {}_2F_1\left[-n,-n';-N;\sin^{-2}\left(\frac{\pi\alpha}{4}\right)\right] = F_{n',n}^{\alpha}
$$
\n(S72)

$$
=e^{i\frac{\pi}{2}(n'-n-N\alpha/2)}\sqrt{\frac{n!\left(N-n\right)!}{n'!\left(N-n'\right)!}}\sin^{n'-n}\left(\frac{\pi\alpha}{4}\right)\cos^{N-n'-n}\left(\frac{\pi\alpha}{4}\right)k_n^{\left[\sin^2\left(\pi\alpha/4\right)\right]}(n',N)\tag{S73}
$$

$$
= e^{i\frac{\pi}{2}(n'-n-N\alpha/2)}\phi_n^{(p)}(n'-Np,N),\tag{S74}
$$

where $k_n^{(p)}(n', N)$ is a Kravchuk polynomial and $\phi_n^{(p)}(n'-Np, N)$ is a Kravchuk function.

We used the following relations^{[25,](#page-42-6)[26](#page-42-7)}

$$
k_n^{(p)}(n',N) = (-1)^n \binom{N}{n} p^n \, {}_2F_1 \left[-n, -n'; -N; \frac{1}{p} \right],\tag{S75}
$$

$$
\phi_n^{(p)}(n'-Np,N) = \sqrt{\frac{n! \ (N-n!)}{n'! \ (N-n)!}} \sqrt{p^{n'-n}(1-p)^{N-n-n'}} \ k_n^{(p)}(n',N),\tag{S76}
$$

$$
\phi_n^{(p)}(n'-Np, N) = (-1)^{n+n'} \phi_{n'}^{(p)}(n-Np, N), \tag{S77}
$$

as well as the fact that the Kravchuk functions are orthonormal

$$
\sum_{n'=0}^{N} \phi_n^{(p)}(n'-Np, N) \phi_m^{(p)}(n'-Np, N) = \delta_{n,m}.
$$
 (S78)

Now we turn $A_S(k, l)$ shown in Eq. [\(S69\)](#page-26-0) to the form of Eq. [\(S74\)](#page-27-0)

$$
\mathcal{A}_{S}(k,l) = e^{-i\theta} \frac{S}{2} e^{i\varphi(l-k)} (-1)^{k+l} \phi_{k}^{(r)}(l - S_{r}, S)
$$

\n
$$
= e^{i\frac{\pi}{2} \left(\frac{2(\pi + \varphi)}{\pi} (l - k) - S_{\pi}^{\theta}\right)} \phi_{k}^{(r)}(l - S_{r}, S)
$$

\n
$$
= e^{-i\theta} \frac{S}{2} e^{i\varphi(l-k)} \phi_{l}^{(r)}(k - S_{r}, S)
$$

\n
$$
= e^{i\frac{\pi}{2} \left(\frac{2\varphi}{\pi} (l - k) - S_{\pi}^{\theta}\right)} \phi_{l}^{(r)}(k - S_{r}, S),
$$
\n(S79)

where $r = \sin^2 \frac{\theta}{2}$.

In specific, if we take $\varphi = -\frac{\pi}{2}$ $\frac{\pi}{2}$ and rearrange terms

$$
\mathcal{A}_{S}(k,l) = e^{i\frac{\pi}{2}(k+l-S\frac{\theta}{\pi})} \sqrt{\binom{S}{k}\binom{S}{l}} \left(\cos\frac{\theta}{2}\right)^{S} \left(\tan\frac{\theta}{2}\right)^{l+k} {}_{2}F_{1}\left[-k,-l;-S;\left(\sin\frac{\theta}{2}\right)^{-2}\right] \quad (S80)
$$

$$
=F_{k,l}^{\frac{20}{\pi}}\tag{S81}
$$

$$
=e^{i\frac{\pi}{2}(l-k-S_{\pi}^{\theta})}\phi_k^{\left(\sin^2\frac{\theta}{2}\right)}(l-S\sin^2\frac{\theta}{2},S),\tag{S82}
$$

$$
\mathcal{A}_S(k,l) = e^{i\frac{\pi}{2}(l-k-S\frac{\theta}{\pi})} \phi_k^{(r)}(l-Sr, S).
$$
\n(S83)

Quantum Kravchuk transform on a beam splitter. Let us send a superposition $\sum_{l=0}^{S} x_l |l, S-l\rangle$ to a BS. The superposition amplitudes encode the sequence (x_1, \ldots, x_S) to be transformed. We will compute the probabilities of detecting $|k\rangle$ and $|S - k\rangle$ photons behind the BS

$$
\left| \langle k, S - k | U_0 U_{BS} \left(\sum_{l=0}^S x_l \cdot | l, S - l \rangle \right) \right|^2 = \left| \sum_{l=0}^S x_l \cdot \langle k, S - k | U_0 U_{BS} | l, S - l \rangle \right|^2 \tag{S84}
$$

$$
= \left| \sum_{l=0}^{S} x_l \cdot e^{-i\theta} \frac{S}{2} \langle k, S - k | U_{BS} | l, S - l \rangle \right|^2 \quad (S85)
$$

$$
= \left| \sum_{l=0}^{S} x_l \cdot \mathcal{A}_S^{(r)}(k,l) \right|^2 \tag{S86}
$$

$$
= \left| \sum_{l=0}^{S} x_l \cdot e^{-i\theta \frac{S}{2}} e^{i\frac{\pi}{2}(l-k)} \phi_k^{(r)}(l - Sr, S) \right|^2 \tag{S87}
$$

$$
= |X_k|^2. \tag{S88}
$$

It is clear now that multi-photon interference on a beam splitter followed by photon-counting detection implements $\alpha = \frac{2\theta}{\pi}$ $\frac{2\theta}{\pi}$ -fractional QKT of the input probability amplitudes

$$
(x_0, x_1, \dots, x_S) \to (|X_0|^2, |X_1|^2, \dots, |X_S|^2),
$$
\n(S89)

where $|X_k|^2$ are experimentally determined photon number statistics for $k = 0, \ldots, S$.

3 Gauss hypergeometric function.

Definition. The Gauss hypergeometric function is a special function defined with the following hypergeometric series

$$
{}_{2}F_{1}(a,b;c;z) = \sum_{k=0}^{\infty} \frac{(a)_{k}(b)_{k}}{(c)_{k}} \frac{z^{k}}{k!},
$$
\n(S90)

where a, b and c are parameters, z is an argument and $(x)_k$ is the Pochhammer symbol

$$
(x)_k = x(x+1)(x+2)\cdots(x+k-1).
$$
 (S91)

In general, all ${}_2F_1$ arguments and the parameter may be complex, $a, b, c, z \in \mathbb{C}$ however, within this note the arguments are always integer, $a, b, c \in \mathbb{Z}$ and the parameter is real, $z \in \mathbb{R}$.

Properties. The Pochhammer symbol can be expressed as a division of factorials

$$
\frac{a!}{(a-k)!} = (-1)^k (-a)_k,
$$
\n(S92)

$$
\frac{a!}{(a+k)!} = \frac{1}{(a+1)_k}.
$$
 (S93)

The form of Eq. $(S90)$ implies that the arguments a and b can be swapped

$$
{}_2F_1(a,b;c;z) = \sum_{k=0}^{\infty} \frac{(a)_k (b)_k}{(c)_k} \frac{z^k}{k!} = \sum_{k=0}^{\infty} \frac{(b)_k (a)_k}{(c)_k} \frac{z^k}{k!} = {}_2F_1(b,a;c;z). \tag{S94}
$$

In case of a negative a or b, the infinite sum in Eq. [\(S90\)](#page-29-3) is truncated because $(x)_k = 0$ if x is a negative integer and $k > -x$. Let us assume that $a < 0$ and $b \ge 0 \vee b < a$. Then, let $m = -a$

$$
{}_{2}F_{1}(-m, b; c; z) = \sum_{k=0}^{\infty} \frac{(-m)_{k} (b)_{k}}{(c)_{k}} \frac{z^{k}}{k!}
$$
 cf. (S92)

$$
= \sum_{k=0}^{m} (-1)^{k} \frac{m!}{k! (m-k)!} \frac{(b)_{k}}{(c)_{k}} z^{k}
$$

$$
= \sum_{k=0}^{m} {m \choose k} (-1)^{k} \frac{(b)_{k}}{(c)_{k}} z^{k}.
$$
 (S95)

Moreover, for the same assumptions as in case of Eq. [\(S95\)](#page-30-0), the following transformation can be used to change z to $1 - z$ [NIST Digital Library of Mathematical Functions, 15.8.7]

$$
{}_2F_1(-m,b;c;z) = \frac{(c-b)_m}{(c)_m} {}_2F_1(-m,b;b-c-m+1;1-z). \tag{S96}
$$

Identities analogous to Eqs. [\(S95\)](#page-30-0) and [\(S96\)](#page-30-1) are also valid for negative b and $a \geq 0 \vee a < b$, due to Eq. [\(S94\)](#page-30-2).

Finally, the following Pfaff's hypergeometric transformation is valid for any a, b, c and z

$$
{}_2F_1(a,b;c;z) = (1-z)^{-b} {}_2F_1(c-a,b;c;z/(z-1)).
$$
\n(S97)

4 Characterization of the setup

In order to estimate transmission losses, we performed Klyshko efficiency measurements on the setup. In a Klyshko measurement with one SPDC source and binary detectors, one counts single events C_A , C_B from either output channel and coincidence clicks C_{AB} between both channels and defines the Klyshko efficiencies η_A and η_B

$$
\eta_B = \frac{C_{AB}}{C_A} \tag{S98}
$$

and vice versa. For low pump powers, these Klyshko efficiencies show a linear pump power dependency, and their intercept is a measure at zero pump power of total transmission efficiency (including both propagation and detection losses) of the associated spatial mode^{[27](#page-42-8)}.

We pumped each of our SPDC sources, one at a time, with the variable beam-splitter in position 50 : 50, at successively lower power values. The resulting four-mode correlated photon statistics were then transformed into binary "photon(s)/no-photon" datasets to emulate standard binary detectors such as avalanche photo-diodes, and we determined the total efficiencies of the heralding modes to be $\eta_1 = 50.3\%$ and $\eta_4 = 48.5\%$. The beam-splitter modes, carrying each a 3 dB loss from the splitter itself and an additional 1 dB due to splitter insertion loss and fiber-tofiber coupling loss, exhibit a total efficiency of $\eta_2 = 21.6\%$ and $\eta_3 = 20.6\%$. Taking into account the additional optical elements in the splitter modes, the efficiencies are consistent. We account for the transmission losses of approximately $50\% \approx 3$ dB with 1 dB initial fiber in-coupling loss due to spatial mode mismatch, 0.25 dB from imperfect detectors, and the rest from three FC/PC fiber-to-fiber couplers per mode as well as bending losses in the transmission fibers between the experimental setup and the detectors.

Fig. [S2](#page-32-0) shows the standard HOM interference dip between both sources measured with binary detectors (InGaAs APDs) for a small mean photon number of the order of 10^{-4} in order to test the setup. The maximal visibility achieved is $V_{HOM} = 85.9\%$. An independent measurement of the second order correlation function for each SPDC source $g^{(2)} = \frac{\langle n^2 \rangle - \langle n \rangle}{\langle n \rangle^2}$ $\frac{\epsilon_{1}-\langle n\rangle}{\langle n\rangle^2} \geq 1.86 \approx 1 + V_{\text{HOM}}$ is consistent with this result. From this, we can infer an effective Schmidt mode number of $K = \frac{1}{a^{(2)}}$ $\frac{1}{g^{(2)}-1} = 1.16$,^{[28](#page-42-9)} i.e. both of our SPDC sources are close to being single-mode.

The TES detectors used in the experiment were thoroughly characterized with quantum to-mography methods^{[29](#page-43-0)}. Their quantum efficiency is above 90%.

Figure S2: HOM dip. Black dots with error bars represent the experimental results whereas the blue line is a fitted curve. The maximal visibility amounts to 85.9% which proves the quantum nature of impinging multiphoton states.

5 Analysis of the experimental data

HOM visibilities. The second-order visibility exceeding the classical value of 50% certifies quantum nature of the HOM interference and thus, the fractional QKT. The visibility is computed with the following formula 30

$$
v^{(2)} = \frac{n_{\text{max}} - n_{\text{min}}}{n_{\text{max}} + n_{\text{min}}},
$$
\n(S99)

where n_{max} and n_{min} are the maximal and minimal number of events registered by the TES detectors for the given photon number S.

The obtained values are gathered in Tab. [1.](#page-33-0) For $S = 5$ it was always greater than 50%. The visibility of interference of $|1, 1\rangle$ given in this Table is much lower than the one reported in Fig. [S2.](#page-32-0) This is because in order to perform quantum simulations with $S > 2$ we increased the power of a laser pumping our source. However, this power increase is too small to affect the analysis from Sec. [4,](#page-30-4) i.e. the source operates in the parametric regime and our photon-number states are near single-mode.

$ \psi\rangle$	$r = 0.05$	$r=0.3$	$r=0.5$	$r = 0.95$
$ 0,1\rangle$	$87.2\% \pm 0.1\%$	$35.1\% \pm 0.1\%$	$0.8\% \pm 0.0\%$	$87.8\% \pm 0.1\%$
	$(\bar{n} = 0.2134)$	$(\bar{n} = 0.2082)$	$(\bar{n} = 0.2097)$	$(\bar{n} = 0.2082)$
$ 0,2\rangle$	$98.2\% \pm 0.3\%$	$59.9\% \pm 0.2\%$	$26.5\% \pm 0.1\%$	$99.0\% \pm 0.3\%$
	$(\bar{n} = 0.2134)$	$(\bar{n} = 0.2082)$	$(\bar{n} = 0.2097)$	$(\bar{n} = 0.2082)$
$ 0,3\rangle$	$99.7\%\pm0.8\%$	$78.6\% \pm 0.7\%$	$52.4\% \pm 0.4\%$	$99.9\% \pm 0.8\%$
	$(\bar{n} = 0.2134)$	$(\bar{n} = 0.2082)$	$(\bar{n} = 0.1996)$	$(\bar{n} = 0.2082)$
$ 0,4\rangle$	$99.1\% \pm 2.5\%$	$87.6\% \pm 2.2\%$	$65.7\% \pm 1.7\%$	$99.9\% \pm 2.5\%$
	$(\bar{n} = 0.2134)$	$(\bar{n} = 0.2082)$	$(\bar{n} = 0.2097)$	$(\bar{n} = 0.2082)$
$ 0, 5\rangle$	$97.8\% \pm 6.2\%$	$96.7\% \pm 7.2\%$	$71.4\% \pm 4.6\%$	$98.6\% \pm 7.2\%$
	$(\bar{n} = 0.2076)$	$(\bar{n} = 0.2082)$	$(\bar{n} = 0.2097)$	$(\bar{n} = 0.1983)$
$ 1,2\rangle$	$74.8\% \pm 0.8\%$	$18.9\% \pm 0.3\%$	$50.3\% \pm 0.2\%$	$79.3 \pm 0.8\%$ %
	$(\bar{n} = 0.2043)$	$(\bar{n} = 0.2082)$	$(\bar{n} = 0.1997)$	$(\bar{n} = 0.2051)$
$ 2,2\rangle$	$94.5\% \pm 2.2\%$	$42.5\% \pm 1.0\%$	$50.6\% \pm 1.2\%$	$93.8\% \pm 2.3\%$
	$(\bar{n} = 0.2088)$	$(\bar{n} = 0.2141)$	$(\bar{n} = 0.2097)$	$(\bar{n} = 0.2051)$
$ 2,3\rangle$	$97.7\% \pm 7.0\%$	$76.6\% \pm 4.7\%$	$54.8\% \pm 3.7\%$	$99.5\% \pm 7.5\%$
	$(\bar{n} = 0.2043)$	$(\bar{n} = 0.2150)$	$(\bar{n} = 0.2097)$	$(\bar{n} = 0.1969)$

Table 1: Second-order interferometric visibilities in HOM interference. The visibility above 50% proves quantum character of the interference. Two-mode Fock states $|\psi\rangle$ impinging on a beam splitter of a variable reflectivity r implement the fractional QKTs. \bar{n} denotes the mean number of interfering photons reached in the experiment.

Computation of probability distributions and estimation of errors. Experimental demonstration of two-mode multi-photon HOM interference requires collecting photon-number statistics, which are then compared with theoretical probability distributions. The statistics result from multiple measurements performed with the setup depicted in Fig. 1b in the main text. The heralding modes (A & D) inform about the input state fed into the variable BS and together with the output modes are measured by highly efficient photon counting TES detectors. Thus, each measurement results in a 4-tuple consisting of the number of photons registered by TES_{1-4} , denoted as (n_1, n_2, n_3, n_4) and corresponding to photon-number states in modes A–D^{[31](#page-43-1)}. In a single run, the SPDC source produces input Fock states consisting of up to approximately 10 photons with probability governed by the pump power (see the *Methods* section in the main text). The detectors register all possible values of $n_i \in [0, 10]$, $i = 1, \ldots, 4$. The automation software stores this data in a database and assigns the number of events to each possible tuple. During a single 400-second run, approx. 10^9 data points are collected.

In order to obtain a photon-number statistics for a given $r = \sin^2 \frac{\theta}{2}$ and input Fock state a post-processing is required. The database is searched for a given pair (n_1, n_4) which determines the two-mode Fock state at the BS input. Then, only records fulfilling the condition $n_1 + n_4 = n_2 + n_3$ are selected as they may correspond to the case of no losses in all paths. For the given (n_1, n_4) the individual probabilities are computed as

$$
p_S(k, n_1 + n_4 - k) = \frac{N(n_1, k, n_1 + n_4 - k, n_4)}{S(n_1, n_4)},
$$

where $N(n_1, n_2, n_3, n_4)$ denotes the number of events of registering the given 4-tuple, $S(n_1, n_4)$ = $\sum_{m=0}^{n_1+n_4} N(n_1, m, n_1 + n_4 - m, n_4)$ is the total number of contributing data points and k as well as $n_1 + n_4 - k$ are the photon numbers registered at the BS outputs. The full probability distribution consists of $n_1 + n_4 + 1$ values for k ranging from 0 to $n_1 + n_4$.

For the TES detectors, due to the overlap between the outcomes associated with neighboring

photon numbers, an $|n\rangle$ state results in a value of $n \pm 1$, where n is registered with probability over 0.9 and the probabilities of $n - 1$ and $n + 1$ are below 0.1 with $p(n - 1) \gg p(n + 1)$. Therefore, the absolute error of a single measurement $\Delta n = \pm 1$. As the computation of probability is based on $S(n_1, n_4)$ data points, the measurement uncertainty equals

$$
\Delta p = \frac{|\Delta n|}{\sqrt{S(n_1, n_4)}} \approx \frac{1}{\sqrt{S(n_1, n_4)}}.
$$

The data post-processing and error estimation was done with a Python script, which prepared input files for the Asymptote plotting software. The probability distributions for an ideal system were computed with Eq. [\(S70\)](#page-26-1). Factorials and binomial coefficients were approximated with the standard lgamma(n) function.

Realistic theoretical model. Actual experimental results (Fig. 3 in the main text) were compared with an enhanced realistic theoretical model which allowed to assess the imperfections of the system. The model includes the following parameters: average photon numbers at the outputs of both SPDCs, strength of the fiber coupling, losses in heralded and interfering modes as well as efficiencies of individual TES detectors.

The computations are done with 6×6 complex matrices, where the indexes 1-2 correspond to heralded modes and 3-4 to the outputs of the variable beam splitter. The indexes 5-6 are responsible for the losses in modes entering the beam splitter, which are modeled by two additional beam splitters which bring the SPDC outputs B and C to interference with the vacuum state. The TES detectors are described by the probability of detecting n_d photons in a Fock state $|n_{\rm in}\rangle$, given by the following formula

$$
p_{\text{TES}}(n_{\text{in}}, n_d, \eta) = \begin{cases} \binom{n_{\text{in}}}{n_d} (1 - \eta)^{n_{\text{in}} - n_d} \eta^{n_d} & \text{if } n_d \le n_{\text{in}},\\ 0 & \text{otherwise}, \end{cases}
$$
 (S100)

where η is the efficiency of the detector, additionally decreased to model imperfections in optical signal transfer (e.g. fiber coupling). The distribution in Eq. [\(S100\)](#page-35-0) well models detectors used in the experiment 29 29 29 .

The numerical program was written in the Java programming language and run on a standard PC. It allows to compute output probability distributions $p_S(k, l)$ for given set of model parameters and given readouts at heralded modes (n_1, n_4) . The computation results were passed to Python scripts which prepared Asymptote data files to be merged with experimental plots. The computations were performed for the same input Fock states as in Fig. 3 in the main text and mean number of photons equal to 0.2. Then, the program was run for various parameters in order to fit the theoretical distributions to the actual experimental data. The results are presented in Fig. [S3.](#page-37-0)

6 Mapping between qudit and interacting spin- $\frac{1}{2}$ -chain quantum computer architectures

Any state of a d-level qudit can be encoded in a chain of d qubits where only one qubit is excited at a time, i.e. using the single excitation basis $|1, 0, \ldots, 0\rangle$, $|0, 1, \ldots, 0\rangle$, etc. The XY Heisenberg model maps the next-neighbor interaction in the chain to the qudit rotation discussed in the main text.

XY model. Let us consider an interacting chain of N qubits governed by the following Hamiltonian

$$
H_{XY} = \sum_{n=1}^{N} \frac{J_n}{2} \left[\sigma_n^x \sigma_{n+1}^x + \sigma_n^y \sigma_{n+1}^y \right],
$$
 (S101)

where σ_n^x , σ_n^y , σ_n^z are the Pauli operators acting on the *n*th qubit and J_n denote couplings between neighboring qubits in the chain.

We first note that a spin- $\frac{N-1}{2}$ particle corresponds to an N-qubit chain with relabeled basis vectors as $|m\rangle$, where $m = -\frac{N-1}{2} + n - 1^{32}$ $m = -\frac{N-1}{2} + n - 1^{32}$ $m = -\frac{N-1}{2} + n - 1^{32}$. The N-qubit Hilbert space is of dimension 2^N . Let us restrict H_{XY} to the N-dimensional single-excitation subspace of this system. This subspace is

Figure S3: Photon number statistics resulting from Fock state $|l, S - l\rangle$ interference. a) $|0, 3\rangle$, b) $|0, 4\rangle$, c) $|0, 5\rangle$, d) $|1, 2\rangle$, e) $|2, 2\rangle$, f) $|2, 3\rangle$. The BS reflectivities are $r = 0.05$ (green), 0.2 (red), 0.5 (blue) and 0.95 (gray). Vertical bars represent theoretical values computed for a realistic system, while dots are values determined in experiment – the probabilities of detecting $|k\rangle$ and $|S - k\rangle$ photons behind the BS. The parameters of computation: mean number of photons generated by SPDC equal to 0.2, TES detection efficiency -0.9 , fiber coupling -0.7 and overall losses in the system – 50%. The states a)-c) encode sequences $(x_0 = 1, x_1 = 0, \ldots, x_S = 0)$, while in d) – (0, 1, 0, 0), c) – (0, 0, 1, 0, 0), d) – (0, 0, 1, 0, 0, 0), respectively. The measured probabilities set their QKTs $(|X_0|^2, |X_1|^2, \dots, |X_S|^2), |X_k|^2 = |\sum_{l=0}^{S} A_S^{(r)}|$ $\int_{S}^{(r)}(k,l)\cdot x_l|^2$ of fractionality $\alpha = 0.28$ (green), 0.60 (red), 1.00 (blue) and 1.72 (gray).

spanned by the basis vectors $|n\rangle$, $n = 1, ..., N$, corresponding to spin configurations in which all spins are "down" apart from just one spin at the vertex n which is "up", i.e. by the eigenstates of the $\sigma_{tot}^z = \sum_i \sigma_i^z$ operator. Then H_{XY} is identical to the Hamiltonian of a spin- $\frac{N-1}{2}$ particle $H = \lambda S_x$, where λ is a constant. Here $J_n = \frac{\lambda}{2}$ $\frac{\lambda}{2}\sqrt{n(N-n)}$. This particular form of J_n allows us to link the XY with the BS interaction. The BS infinitesimal evolution turn the input state $|l, S - l\rangle$ into the superposition

$$
H_{BS}|l, S-l\rangle = q_{l,l-1}|l-1, S-l+1\rangle + q_{l,l+1}|l+1, S-l-1\rangle, \qquad (S102)
$$

with the amplitudes

$$
q_{l,l+1} = \frac{\sqrt{(l+1)(S-l)}}{2}.
$$
\n(S103)

The amplitudes reproduce J_n for $N = S + 1$, $n = l + 1$ and $\lambda = 1$.

Example: quantum annealing processor. A Hamiltonian describing quantum annealing processor based on N interacting qubits reads

$$
H_S(s) = \mathcal{E}(s)H_P - \frac{1}{2}\sum_i \Delta(s)\sigma_i^x, \quad i = 1, \dots, N,
$$
\n(S104)

where s denotes time (s = t/t_f, t ∈ [0, t_f]), $\mathcal{E}(s)$ and $\Delta(s)$ are the transverse and longitudinal energies, respectively. H_P is a dimensionless Hamiltonian

$$
H_P = -\sum_i h_i \sigma_i^z + \sum_{i < j} J_{ij} \sigma_i^z \sigma_j^z,\tag{S105}
$$

where biases h_i and couplings J_{ij} encode a particular optimization problem. Quantum annealing starts with setting $\Delta \gg \mathcal{E}$, then Δ is reduced and \mathcal{E} is increased until $\mathcal{E} \gg \Delta$ and $H_S \approx H_P$.

Thus, initially the qubit register is prepared in an eigenstate of the $\sigma_{tot}^x = \sum_i \sigma_i^x$ operator and then the following evolution $\sum_{i \le j} J_{ij} \sigma_i^z \sigma_j^z$ is applied (for simplicity we assume $h_i = 0$). If we now take $J_{ij} = J_n$ for two neighboring qubits and $J_{ij} = 0$ otherwise, we will reproduce the evolution in the XY model, where the register is initially in the eigenstate of σ_{tot}^z and evolution takes place in the orthogonal subspace \sum_n J_n 2 $\left[\sigma_n^x\,\sigma_{n+1}^x+\sigma_n^y\,\sigma_{n+1}^y\right]\!.$

How to perform the QKT of MRI data? The MRI frequency data form a matrix of complex coefficients $\{f_{x,y}\}\$, $x, y = 1, \ldots, N$, and their processing requires a two-dimensional QKT. Thus, the input data have to be encoded in a 2D quantum superposition with $\{f_{x,y}\}$ defining its amplitudes For a spin chain implementation this could be the following encoding

$$
|\Psi_{in}\rangle = \sum_{i,j=1}^{N} f_{i,j} |0_1, \dots, 1_i, \dots, 0_N\rangle |0_1, \dots, 1_j, \dots, 0_N\rangle.
$$
 (S106)

Here a long chain of spins is divided into two subchains, and the operations on them are performed independently.

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