

Applications of Quantum Channel Discrimination

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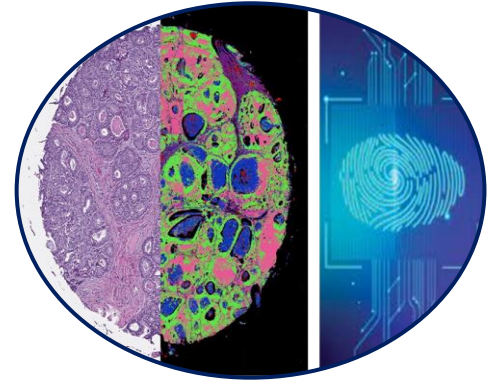
Department of Physics and Astronomy- Unifi (Florence)



Quantum metrology and Hypothesis testing

Beating the **classical limits** in the **estimation/discrimination** of a **parameter** exploiting **quantum resources**

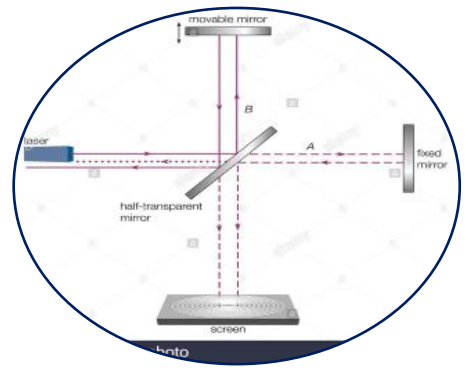
Imaging and pattern recognition



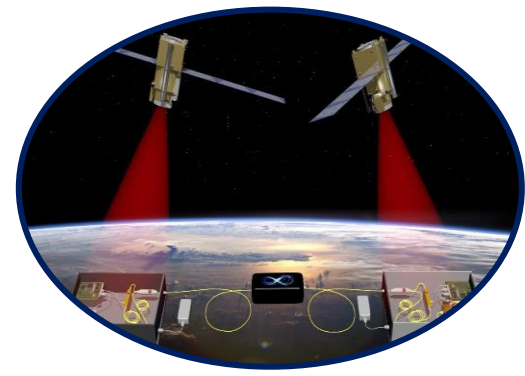
Optical memories readout



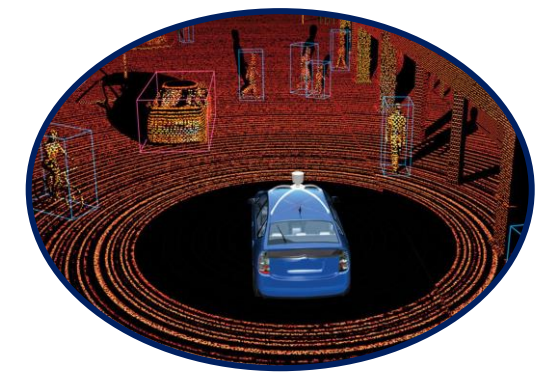
Interferometry



Characterization of quantum technologies

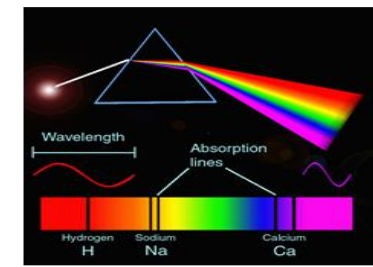
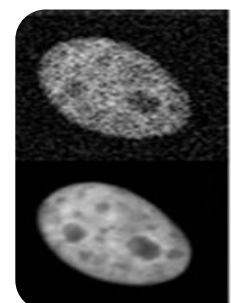
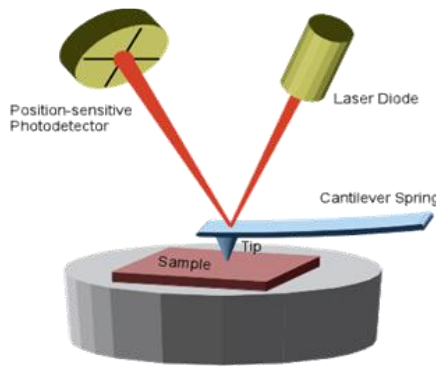
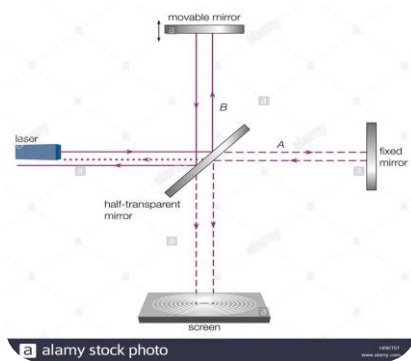


Target detection and ranging



Classical limit: The Shot Noise Limit

Shot Noise Limit (SNL)/Standard Quantum Limit (SQL) comes from the discrete nature of the light and sets the lower bound to the sensitivity of conventional (classical) optical measurements



phase (interferometry)

$$\Delta\varphi_{\min} \approx \frac{1}{\sqrt{N}}$$

Displacement/spatial resolution

$$\Delta x_{\min} \approx \frac{X_R}{\sqrt{N}}$$

Loss/transmission

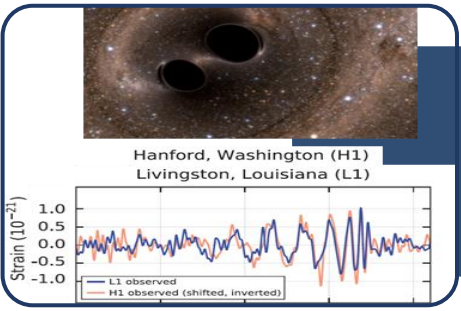
$$\Delta\alpha_{\min} \approx \frac{\sqrt{1-\alpha}}{\sqrt{N}}$$

$0 < \alpha < 1 = \text{loss}$



Classical limit: The Shot Noise Limit

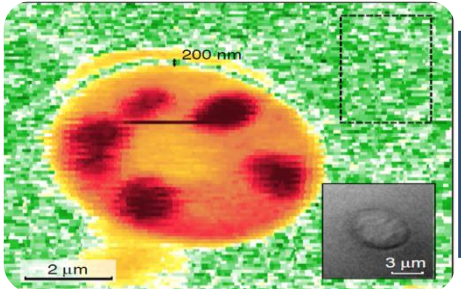
But increasing optical power is not always an option...



Super accurate measurements -> technical noise, quantum back action (GW detection)

$N = 10^{23}$ ph/s (100 kW)

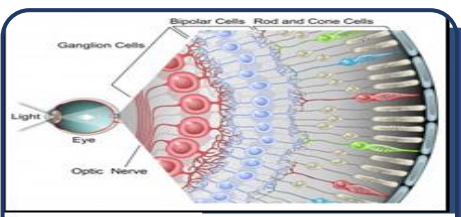
J. Aasi et al., Nat. Photonics 7, 613 (2013)



Probing delicate systems (biological samples, photosensitive chemicals)

$N < 10^{12}$ ph/s (< 1 μW at the sample)

Casacio et al. Nature, 594, 201-205, (2021)



Investigate response of a system at few photons level (SPD, photo-transduction, photosynthesis...)

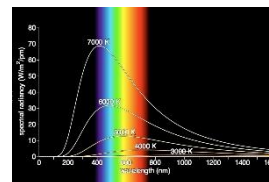
sensitivity threshld $N = 1$ ph!

J.N. Tinsley et al, Nature Comm. 7:12172 (2016)

© Image courtesy of Jamie Simon, Salk Institute.



Classical limit: The Shot Noise Limit



SNL coincides with the uncertainty
statistical scaling of N independent
repetitions (CENTRAL LIMIT
THEOREM)



Classical limit: The Shot Noise Limit



SNL coincides with the uncertainty statistical scaling of N independent repetitions (CENTRAL LIMIT THEOREM)

Quantum correlation: a degree of cooperation among particles which is not possible in the classical description

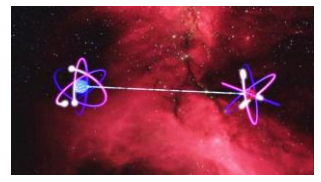


Extra information available in quantum states enables reducing the uncertainty

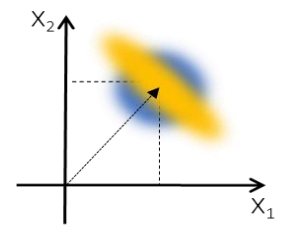


Overcoming the SNL: Quantum resources

- **Entanglement**

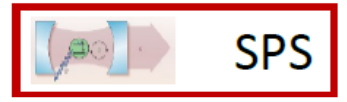


- **Squeezing**

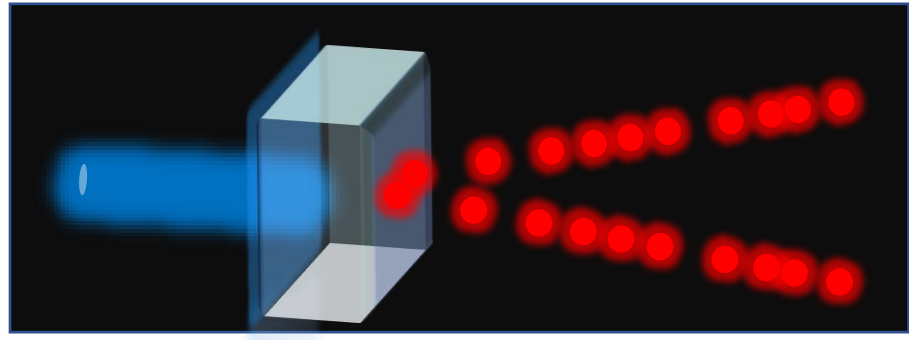


- **Sub-Poisson Light:**

- **Single photon sources**



• **Photon number Non-Classical correlation**

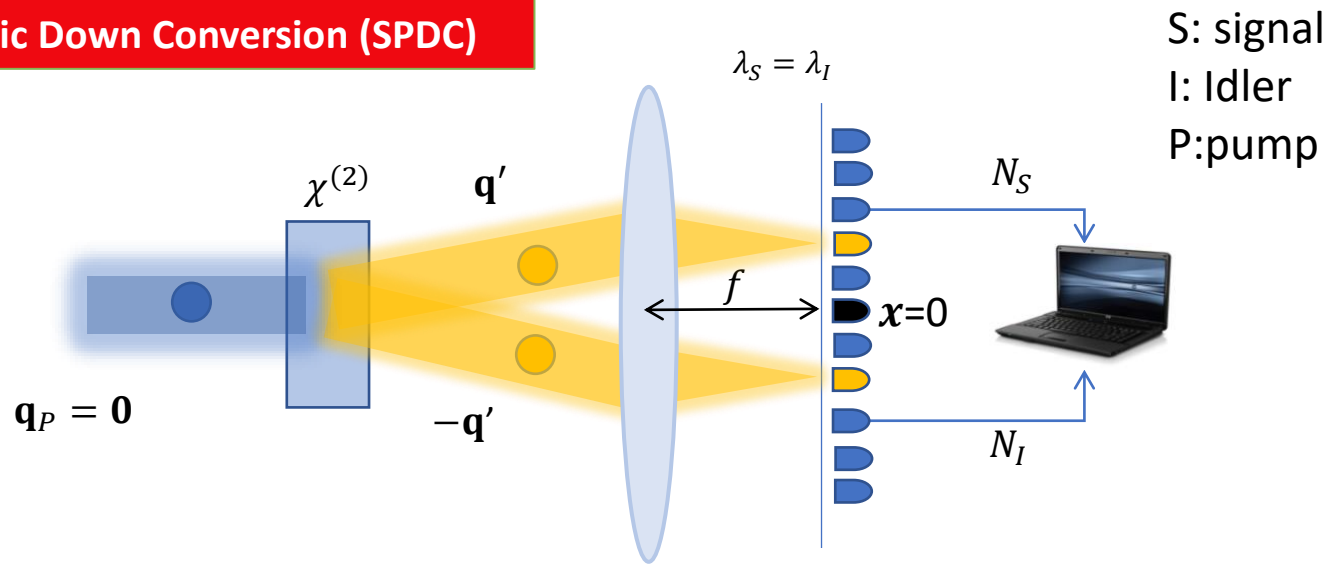


Non-Classical PN correlations: SPDC

Spontaneous Parametric Down Conversion (SPDC)

Energy and momentum conservation

$$\omega_S + \omega_I = \omega_p$$

$$\mathbf{k}_P = \mathbf{k}_S + \mathbf{k}_I$$


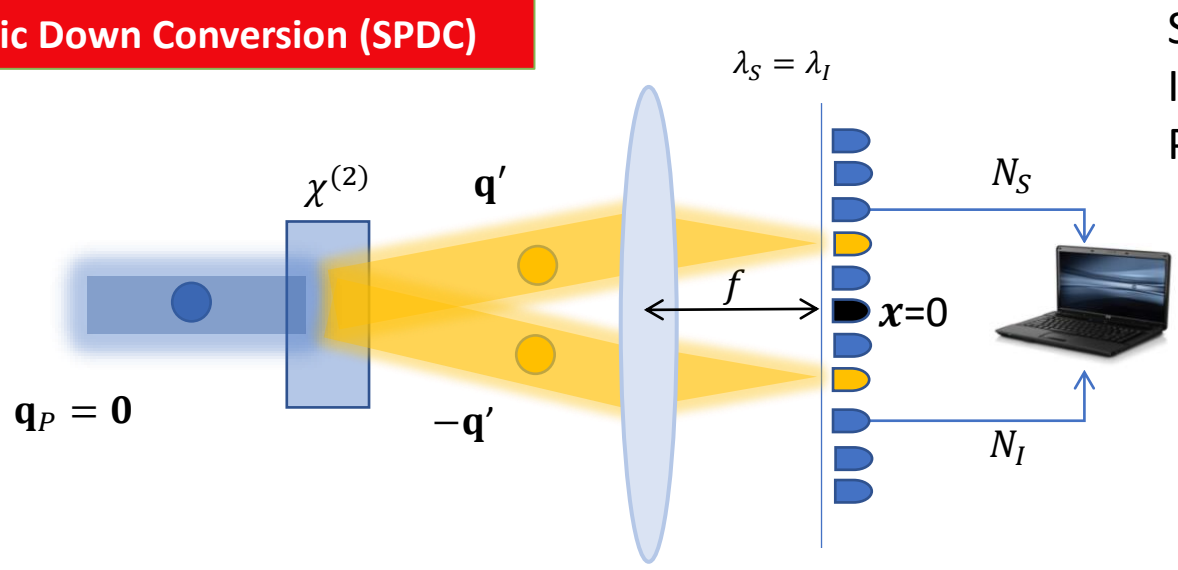


Non-Classical PN correlations: SPDC

Spontaneous Parametric Down Conversion (SPDC)

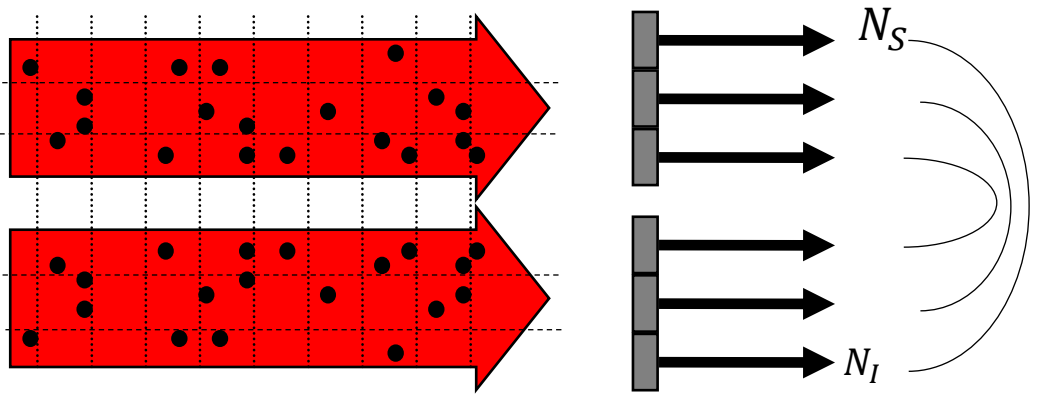
Energy and momentum conservation

$$\omega_S + \omega_I = \omega_p$$

$$\mathbf{k}_P = \mathbf{k}_S + \mathbf{k}_I$$


S: signal
I: Idler
P: pump

Strong **temporal** and **spatial** photon number correlation



Two-Mode Squeezed Vacuum (TMSV) state

$$|\text{TMSV}\rangle_{S,I} \propto \sum_n \sqrt{P_{\mu_0}(n)} |n\rangle_S |n\rangle_I$$

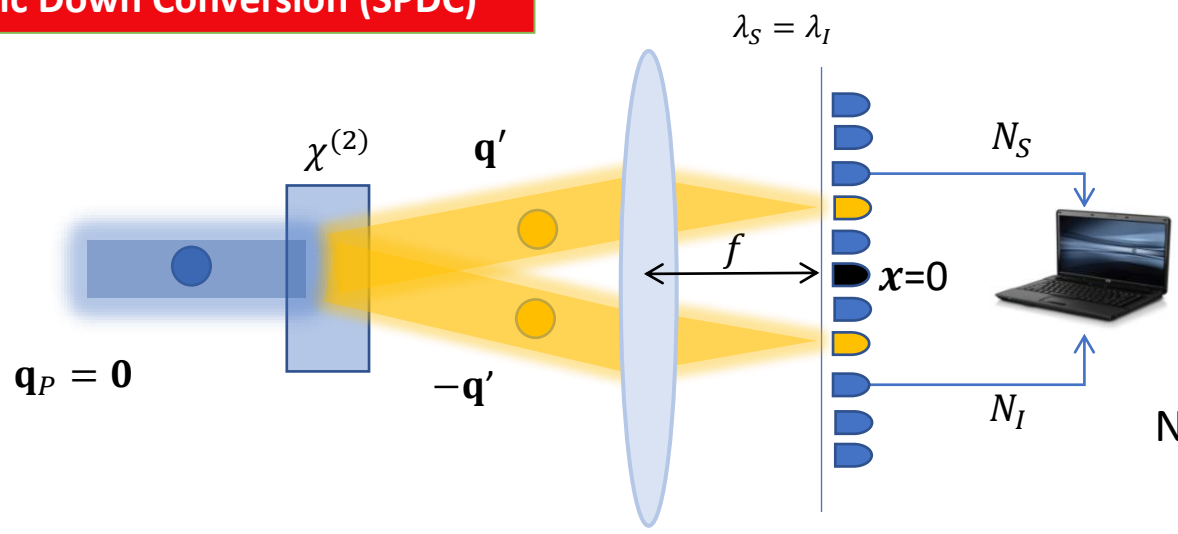
$$P_{\mu_0}(n) = \mu_0^n / (\mu_0 + 1)^{n+1}$$

Non-Classical PN correlations: SPDC

Spontaneous Parametric Down Conversion (SPDC)

Energy and momentum conservation

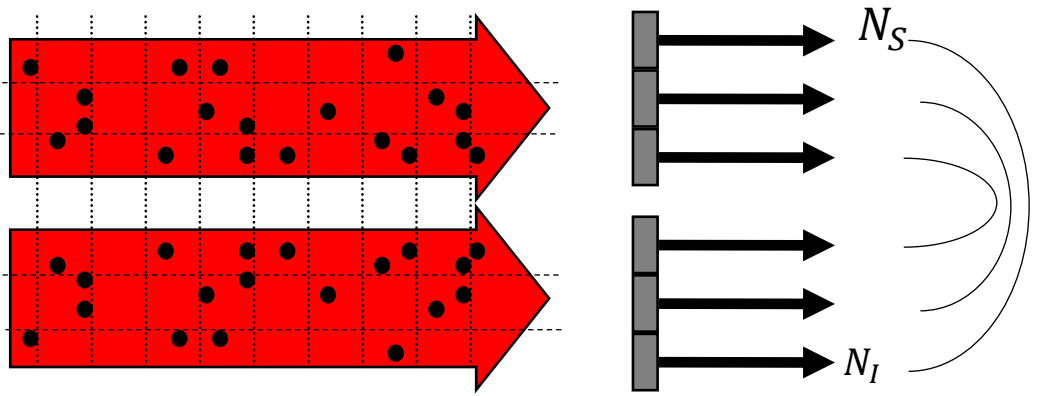
$$\omega_S + \omega_I = \omega_p$$

$$\mathbf{k}_p = \mathbf{k}_S + \mathbf{k}_I$$


η : efficiency in detecting correlated photons

Noise Reduction Factor (NRF) =
$$\frac{\text{Var}(N_S - N_I)}{\langle N_S + N_I \rangle}$$

Strong **temporal** and **spatial** photon number correlation



Two-Mode Squeezed Vacuum (TMSV) state

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$$P_{\mu_0}(n) = \mu_0^n / (\mu_0 + 1)^{n+1}$$

$\text{NRF} \geq 1$ For classical light

$0 \leq \text{NRF} = 1 - \eta < 1$ For TMSV



NON-CLASSICAL



Quantum Hypothesis Testing

Binary state discrimination

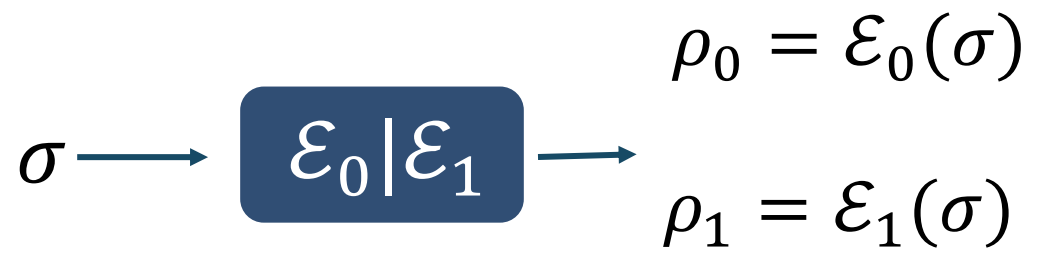
$$\begin{aligned} \mathcal{H}_0: \rho &= \rho_0 \\ \mathcal{H}_1: \rho &= \rho_1 \end{aligned}$$



Helstrom formula

$$p_{err} = \frac{1}{2}(1 - \|\pi_1\rho_1 - \pi_0\rho_0\|)$$

Quantum channel discrimination



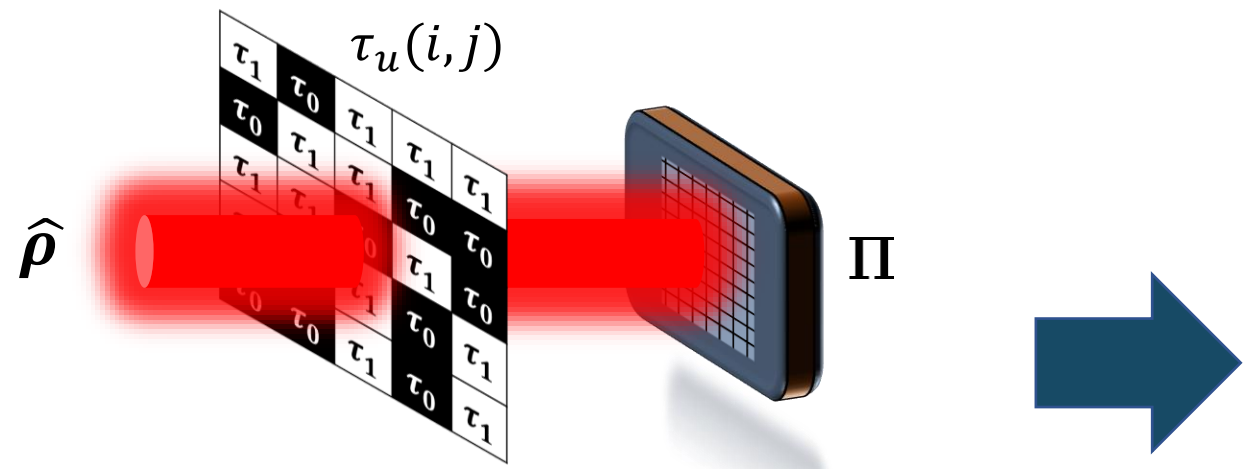
Discrimination of losses

$$\mathcal{E}_{\tau_0}(\sigma) \text{ OR } \mathcal{E}_{\tau_1}(\sigma)$$

Discrimination of bosonic losses

Discrimination of bosonic loss channels

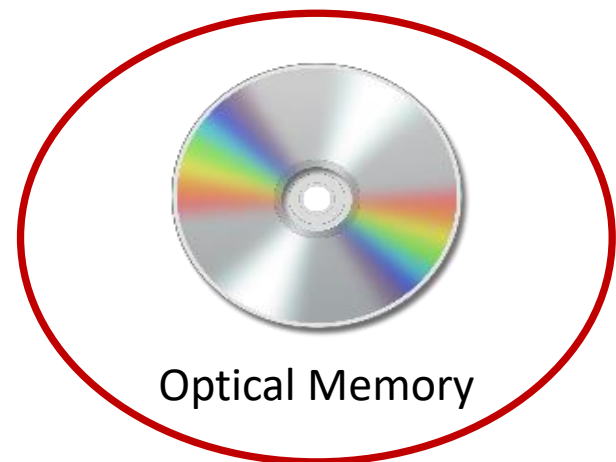
Some Applications



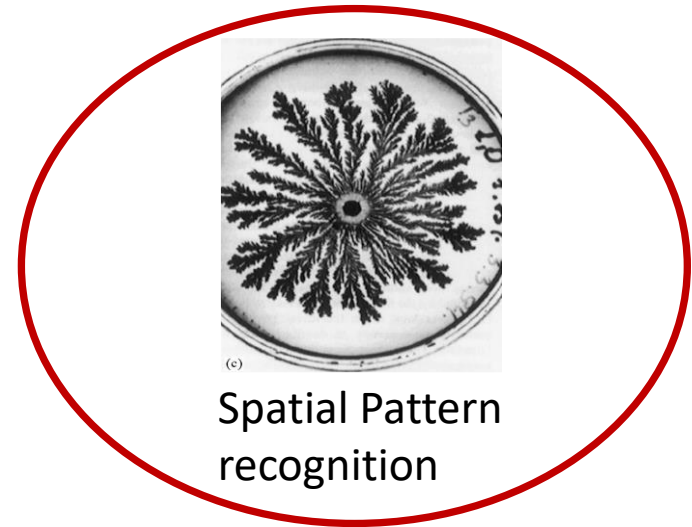
- Each cell (i, j) encodes a bit in two values of optical parameter τ_u ($u = 0, 1$)
- Loss channel discrimination

$v(i, j)$
Assigned bit value

$p_{err}(\hat{\rho}, \Pi)$
Error probability



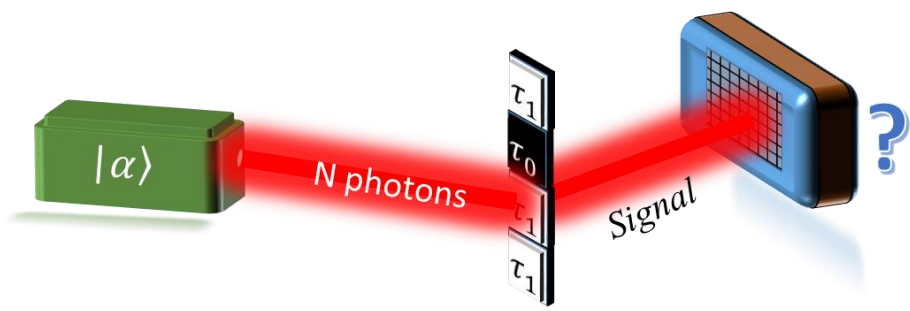
Optical Memory



Spatial Pattern recognition

Quantum reading of a digital memory

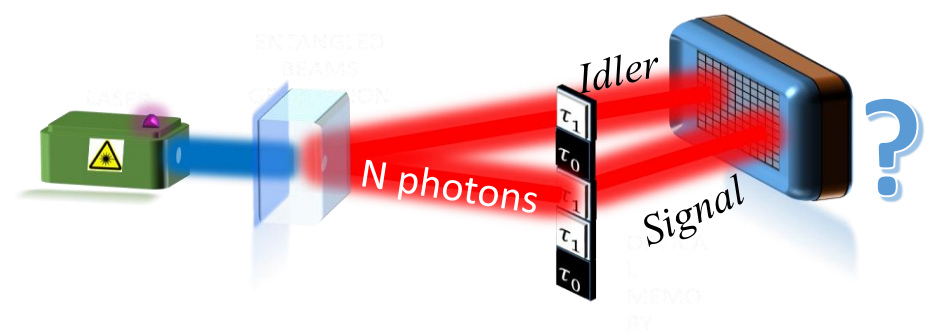
Optimal Classical bound



$$p_{err}^{cla} \geq \mathcal{C}(N, \tau_0, \tau_1) := \frac{1 - \sqrt{1 - e^{-N(\sqrt{\tau_1} - \sqrt{\tau_0})^2}}}{2}$$

- Single coherent mode is the optimal classical transmitter $\otimes^M \rho_{SI} \rightarrow |\alpha\rangle_S$, $|\alpha|^2 = N$ (no ancillary modes needed)
- The optimal measurement is highly theoretical

A specific Quantum Transmitter



M replicas of Two Mode Squeezed Vacuum (TMSV)

$$|TMSV\rangle_{S,I}^{\otimes M} \quad |TMSV\rangle = \sum_{n=0}^{\infty} c_n |n\rangle_S |n\rangle_I$$

Given $N > N_{th}(\tau_0, \tau_1)$ There is \bar{M}

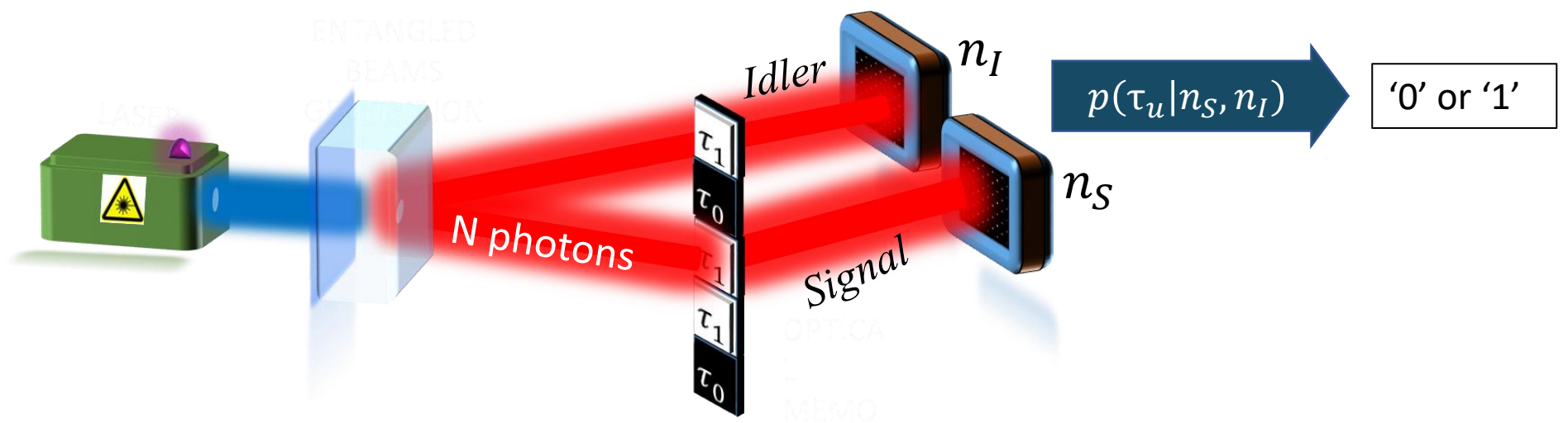
$$p_{err}^{TMSV} < \mathcal{C}(N, \tau_0, \tau_1)$$

Theoretical Quantum advantage



Experimental quantum reading: Photon counting receiver

Practical Implementation: Photon Counting Receiver



M replicas of TMSV

$$|\text{TMSV}\rangle_{S,I}^{\otimes M}$$

+

Local photon counting

$$n_S, n_I$$

+

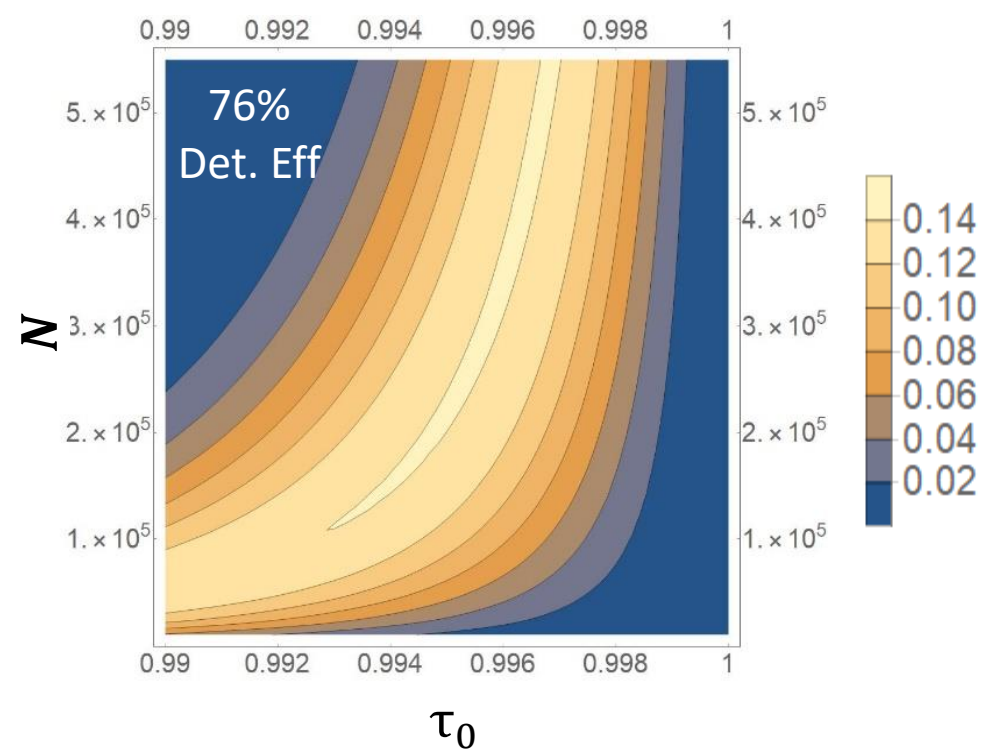
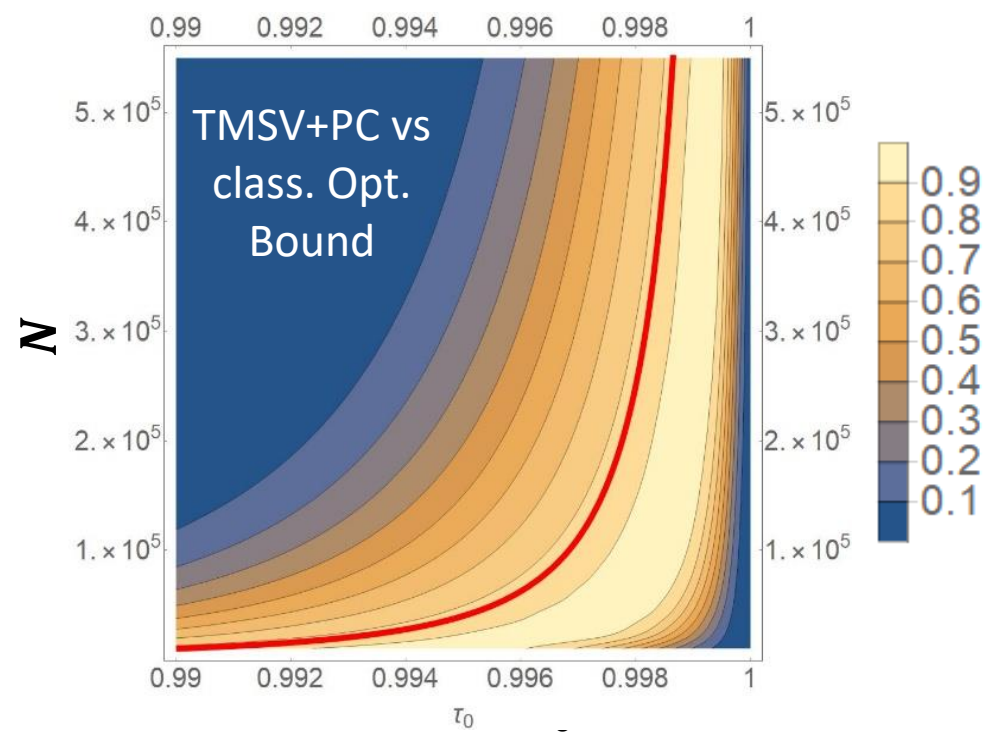
Best Bayesian decision

$$u = \text{argmax}_u p(\tau_u | n_S, n_I)$$

Performance is close to the one theoretically predicted for the optimal receiver

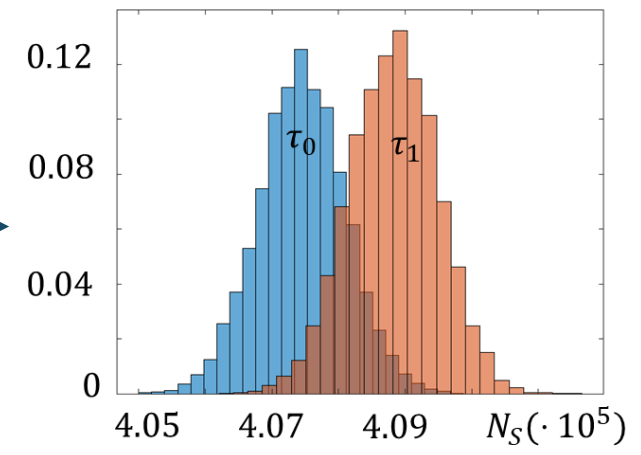
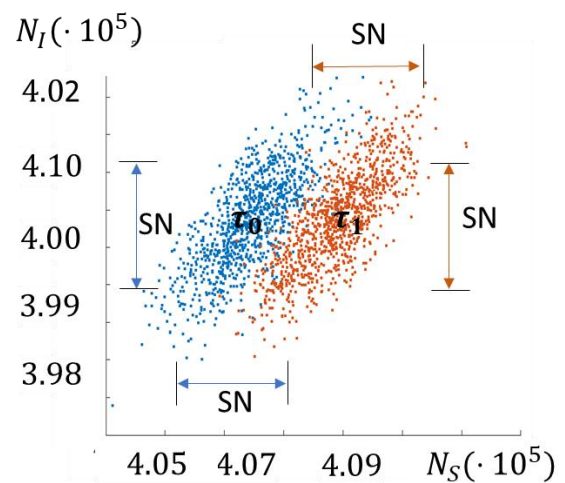
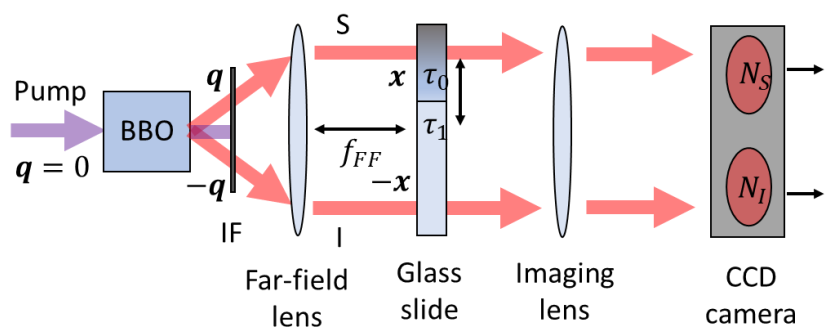
Experimental quantum reading: Results

Theoretical Quantum Gain (bits/cell) in function of transmissivity τ_0 mean number of photons N (higher transmissivity is set to $\tau_1 = 1$)



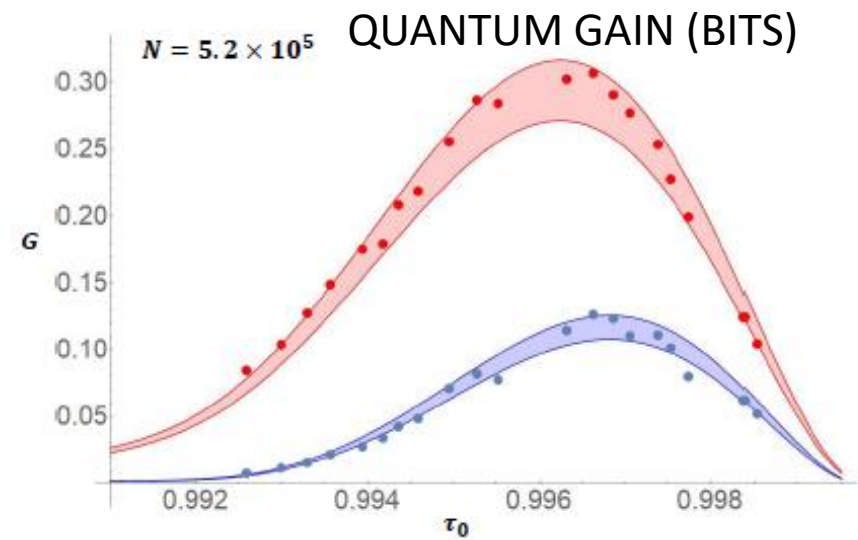
For certain region of the parameter space QR retrieves almost all the information while the best classical strategy completely fails!

Experimental quantum reading: Results



- Single measurement Time: 10 ms
- Spatio-temporal mode: $M \sim 10^{13}$
- Mean number of Photon: $N \sim 10^5$
- Transmittance (bit): $\tau_0 \sim 0.995, \tau_1 = 1$
- Detection efficiency: $\eta_S, \eta_I = 0.78$

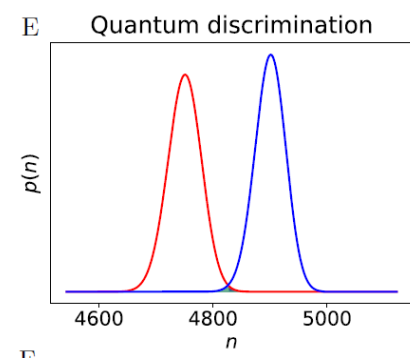
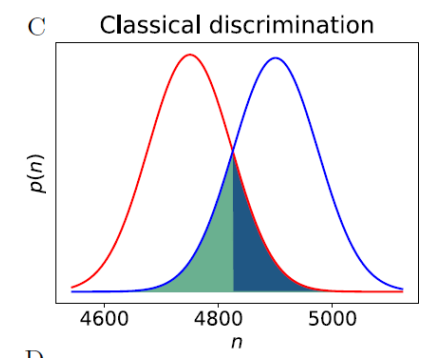
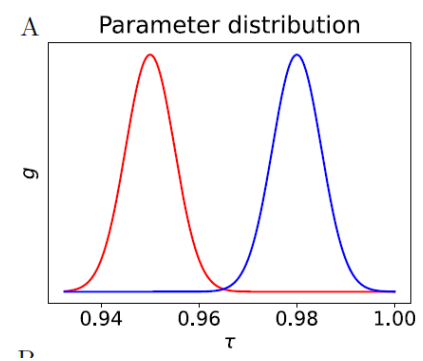
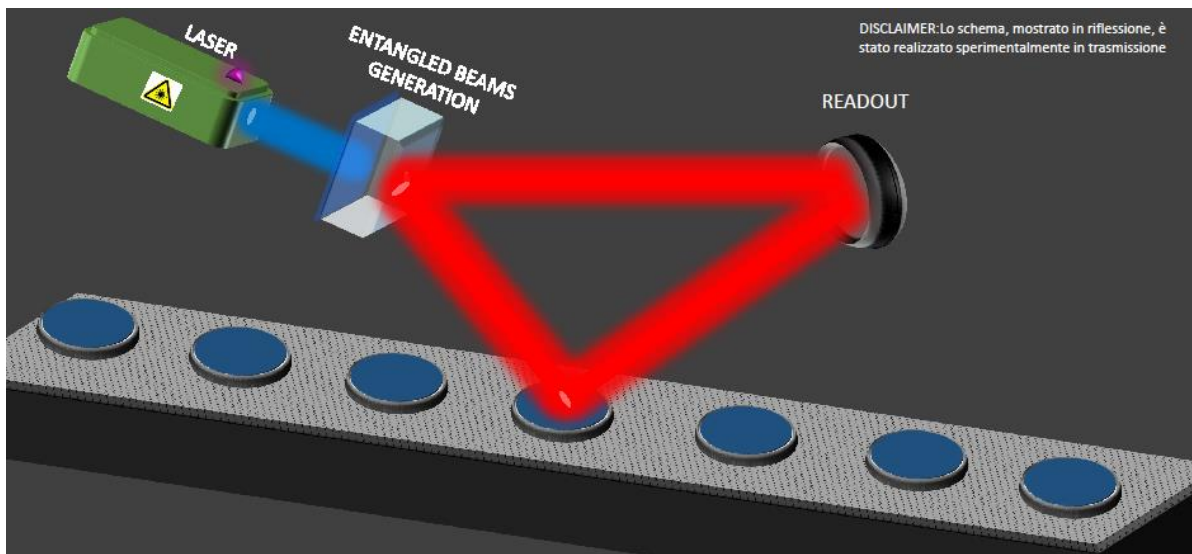
$N/M \ll 1$
Poissonian
marginal distribution
(shot noise limited)



- Gain up to 0.14 bit per cell compared to the optimal classical strategy (coherent state + unknown receiver)
- 0.3 bit per cell compared to a the optimal classical strategy based on photon counting

Distinguishing two distributions of losses

- Production process monitoring (Quantum conformance test)
- Readout of imperfect memories

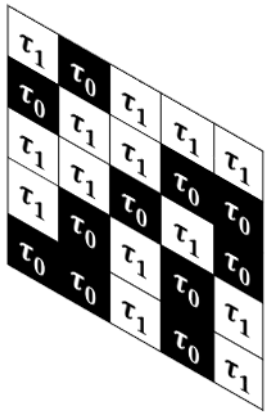


■ False positive
■ False negative

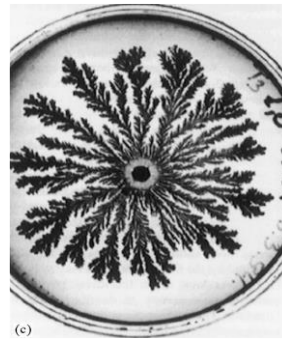
Quantum-enhanced Pattern recognition

From quantum reading/single cell task....

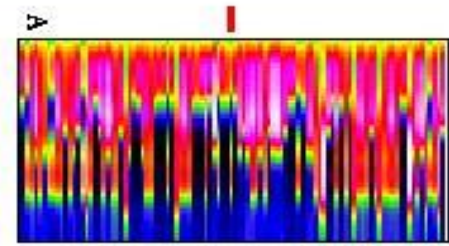
.... To pattern recognition/multicell



Optical Memory



Pattern recognition, e.g. biological structures (Petri Pattern)



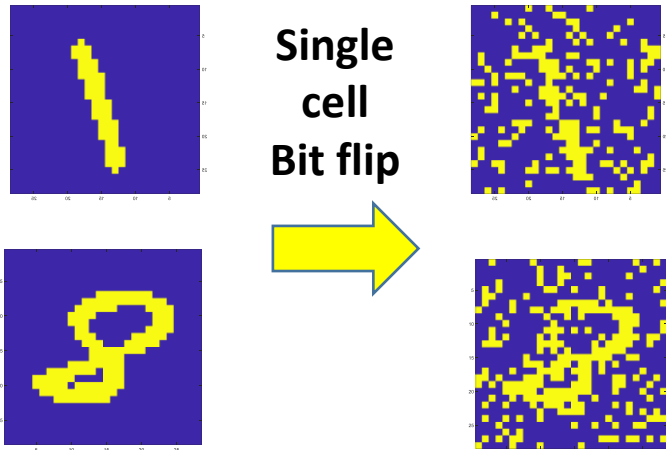
Fingerprints of a substance in spectroscopy

- The relevant information are global features (patterns) rather than just the sum of the individual cells → the task is the pattern classification/recognition

• Is the quantum advantage in sensing preserved when detecting global feature with complex post processing (e.g. machine learning)?

Quantum-enhanced Pattern recognition

Classification of handwritten digits



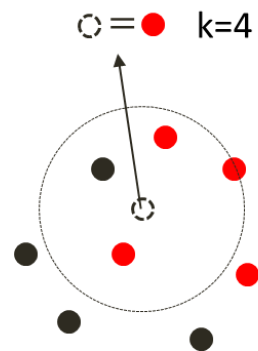
From pixel error to classification error:

$$p_{err}^C = \mathcal{M}_C(p_{err}^{px})$$

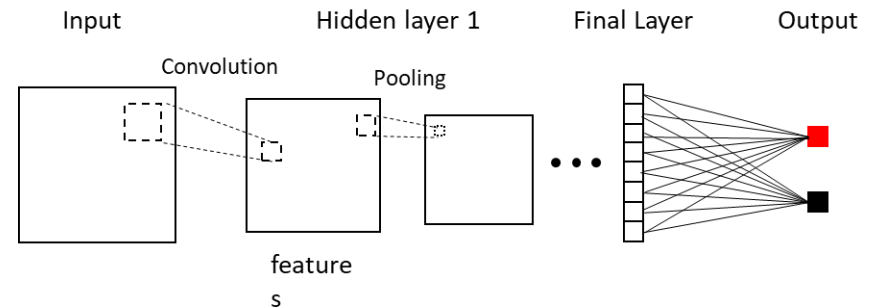
Classification performed with two classification algorithms:

Supervised Classification

k-Nearest Neighbor (K-NN)

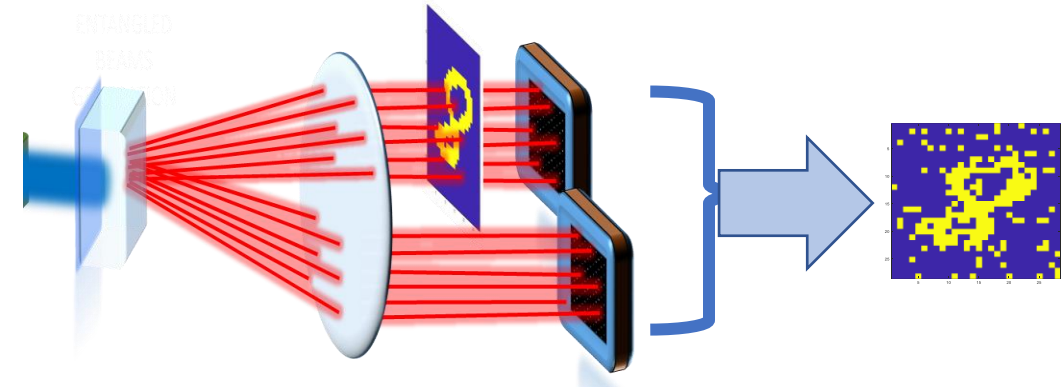


Convolutional Neural Network (CNN)



Quantum-enhanced Pattern recognition

- The photon count in each pixel is used to assign the value of the bit → the final image is binary (black and white)



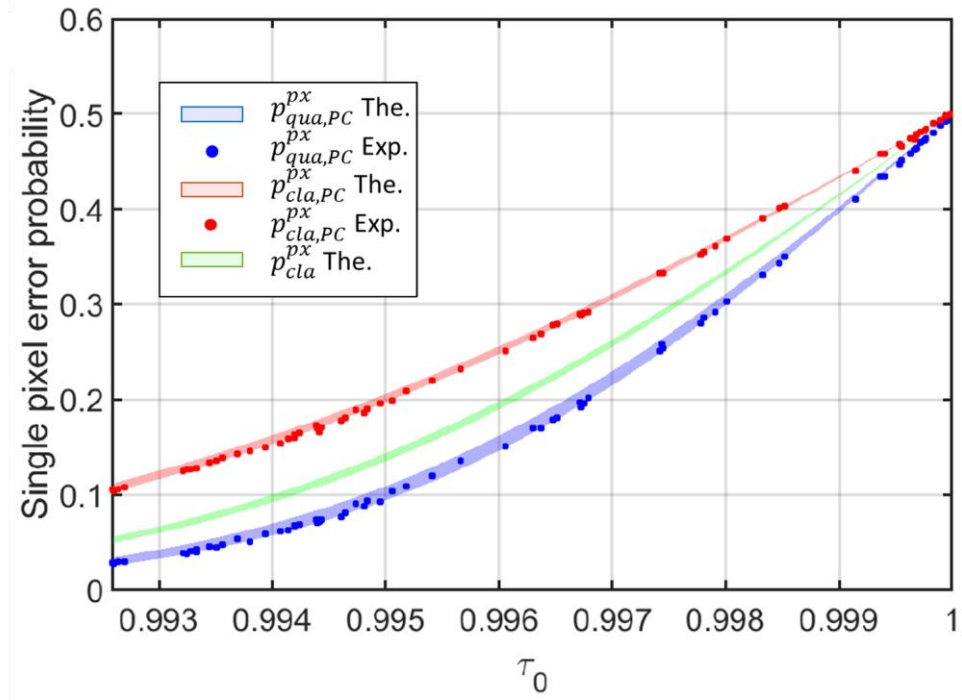
- Binary image are classified by the **supervised NN learning algorithm**

trained on a training set (60000 images)

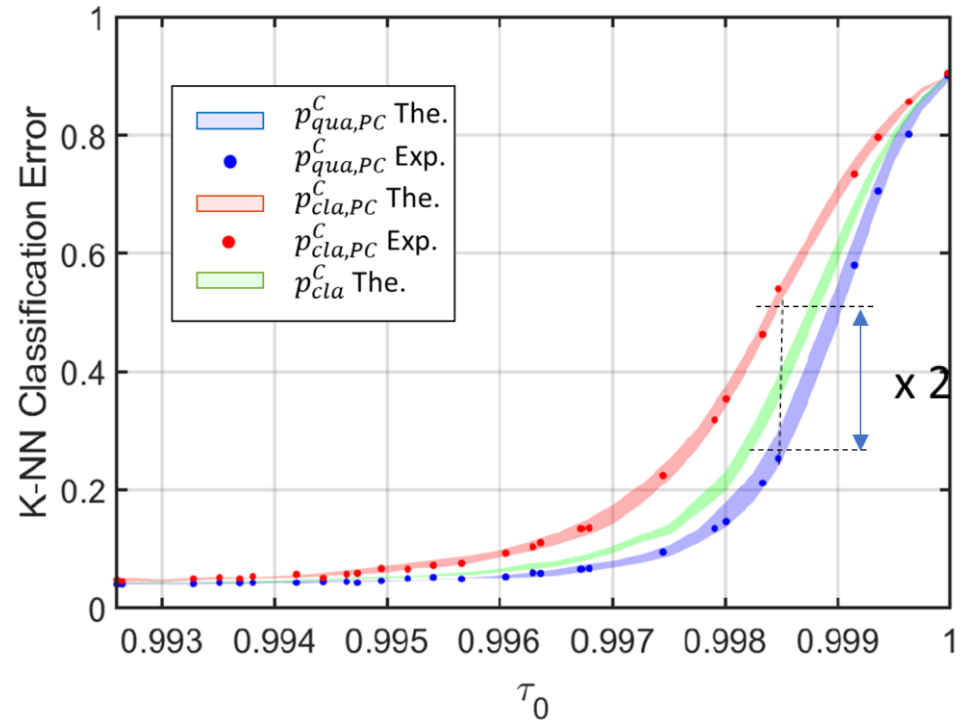




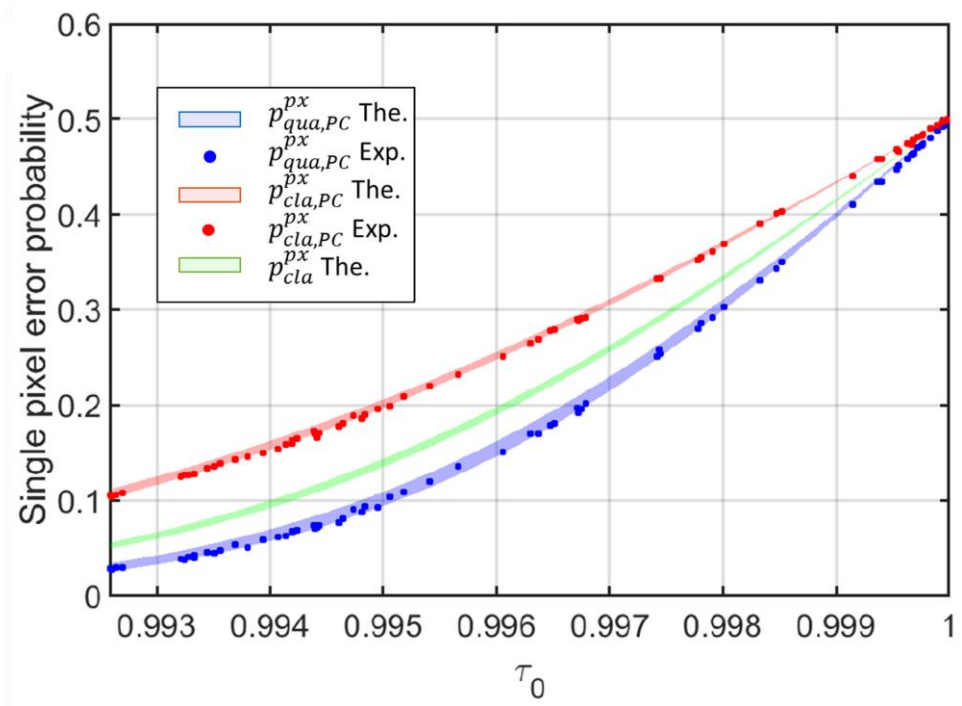
Discrimination of bosonic losses



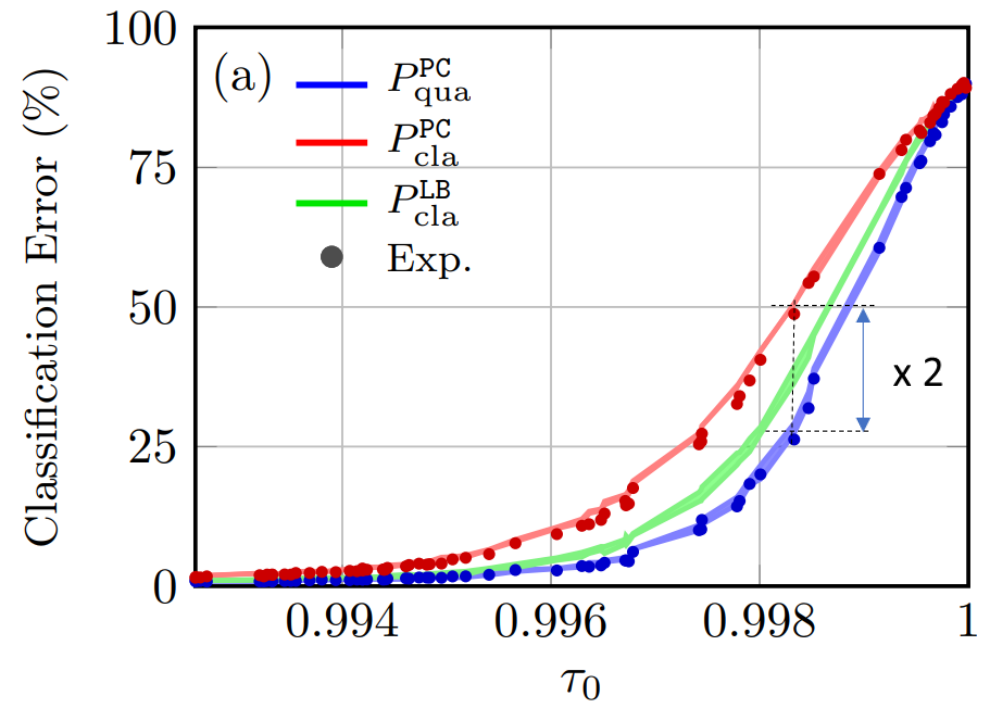
K-Nearest Neighbor (k-NN) classifier



Discrimination of bosonic losses



Convolutional neural network (CNN)



- The CNN classifier behaves very similarly to the K-NN algorithm confirming that the quantum advantage is sustained through sophisticated classical post-processing



Conclusions

- Quantum resources can boost the performance over classical schemes in the optical domain
- Experimental quantum advantage in the readout of a digital memory and related protocols
- Scalability of the sensing advantage to complex tasks such as pattern recognition



Conclusions and perspective

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 - Experimental quantum advantage in the readout of a digital memory and related protocols
 - Scalability of the sensing advantage to complex tasks such as pattern recognition
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- Broader scalability study using **Information Bottleneck**



Conclusions and perspective

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 - Experimental quantum advantage in the readout of a digital memory and related protocols
 - Scalability of the sensing advantage to complex tasks such as pattern recognition
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Thank You!