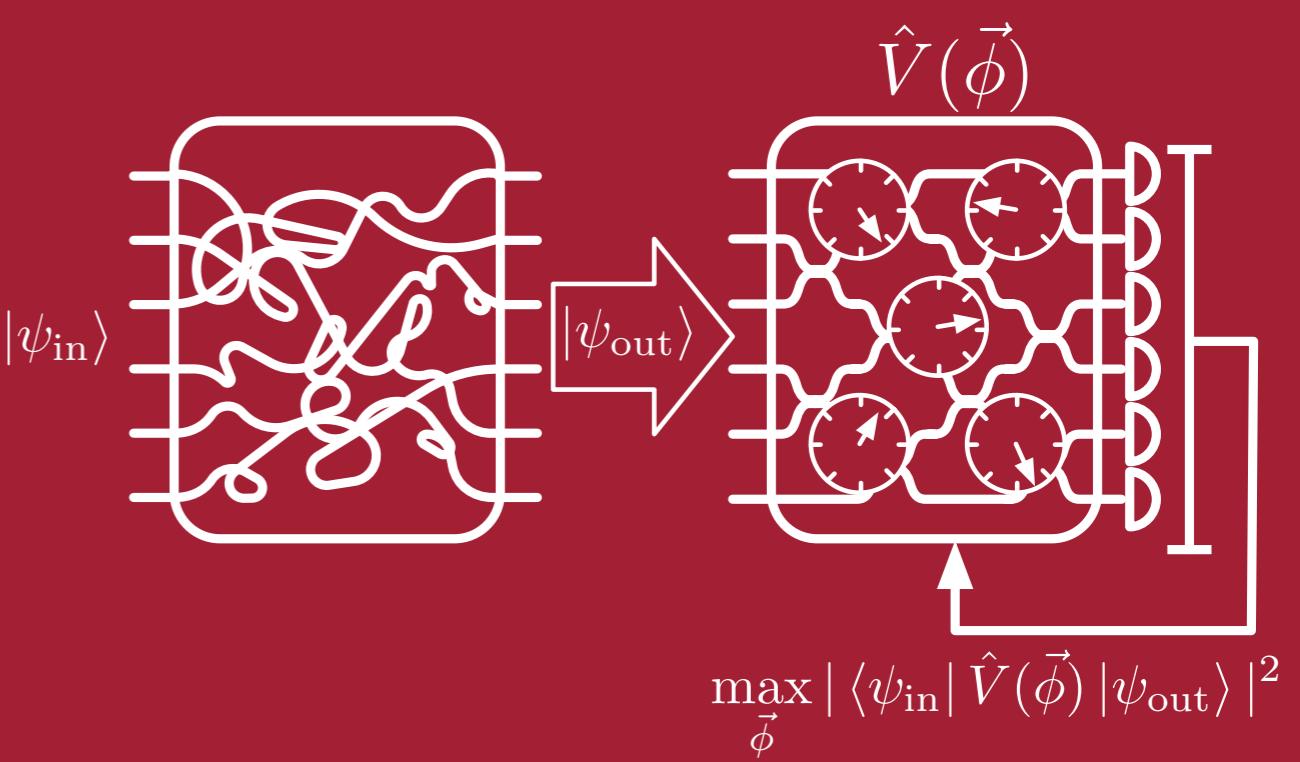
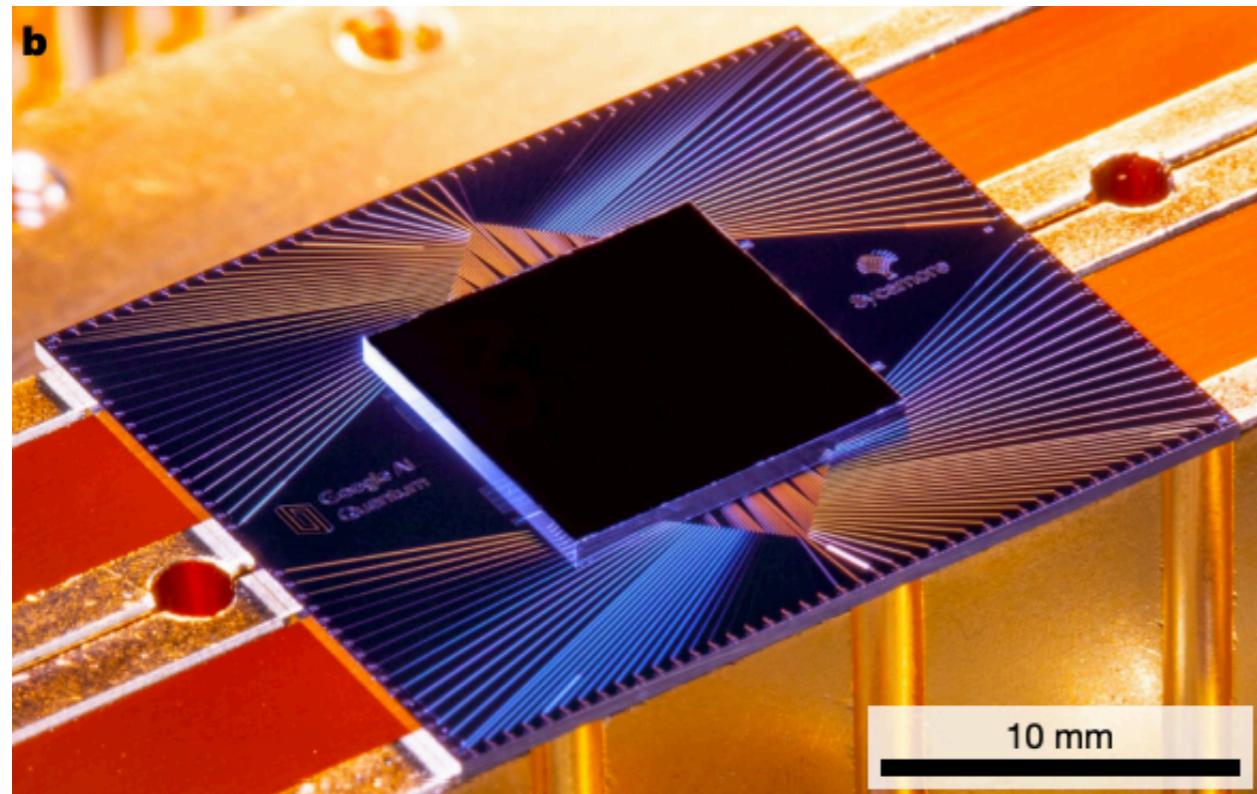


Quantum Photonic Processors to Accelerate Machine Learning

Jacques Carolan, NBI
 @JacquesCarolan



Quantum is Mainstream!



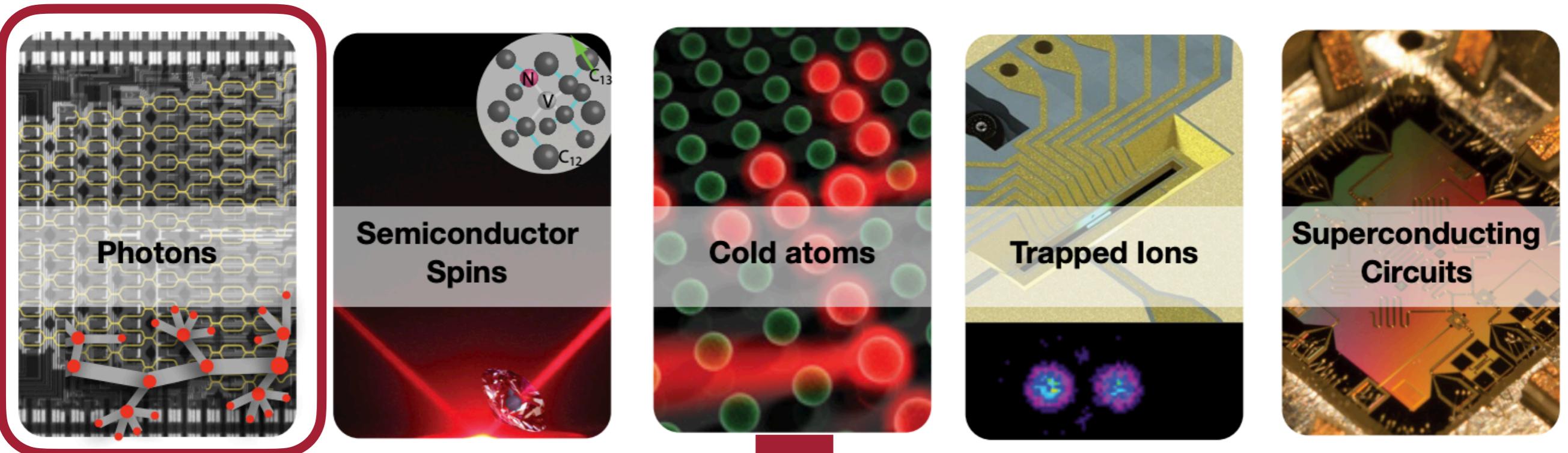
F. Arute et al., *Nature* **574**, 505–510 (2019)

The New York Times

 Google AI

Google Claims a Quantum Breakthrough That Could Change Computing

Qubit Technology



>\$500M



| QuEra >

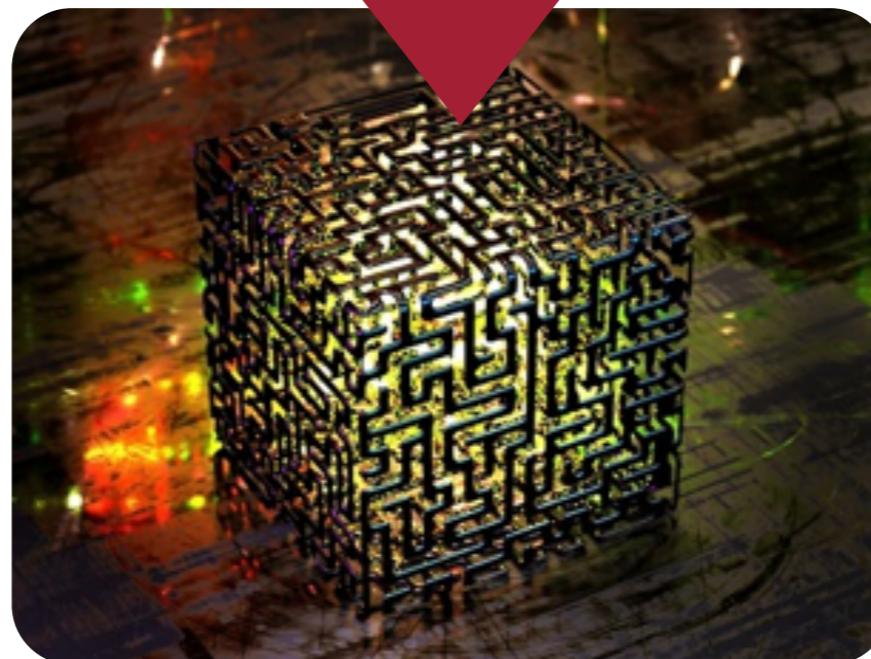


Google AI

IBM Q™



??

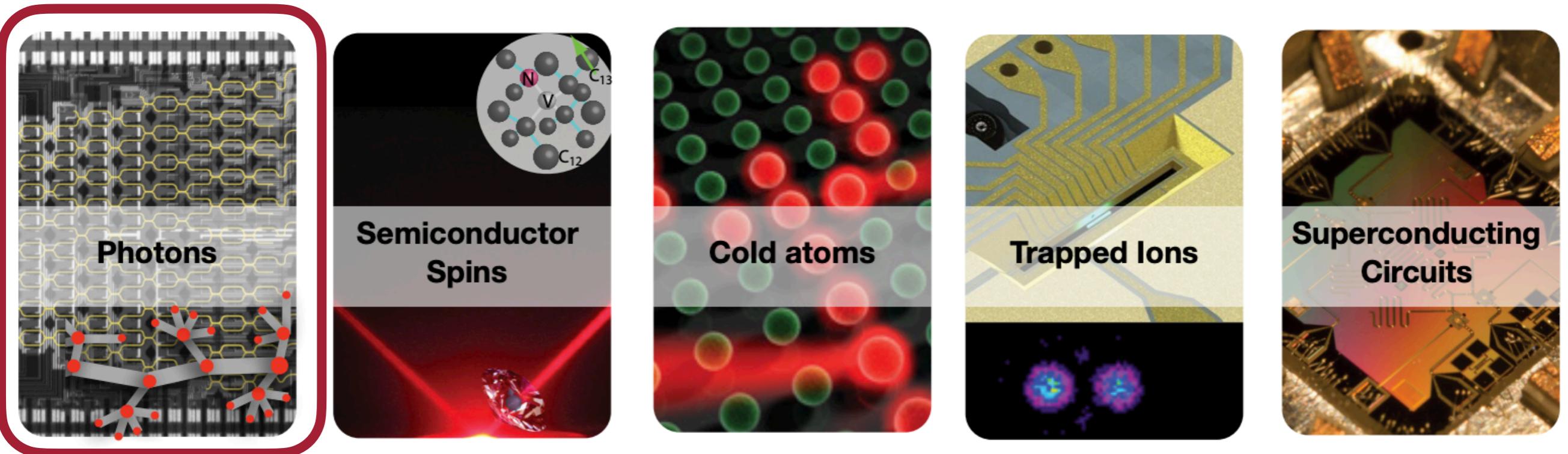


Universal Quantum Computer

Open Questions

- Material science?
- Device design?
- Architectures?
- Engineering at scale?
- Interconnects?
- Errors?
- Useful algorithms?

Qubit Technology



1) Quantum Interconnects

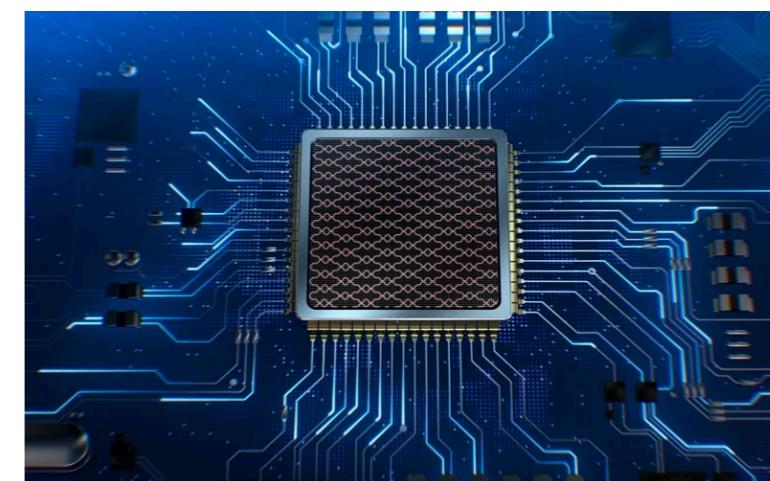
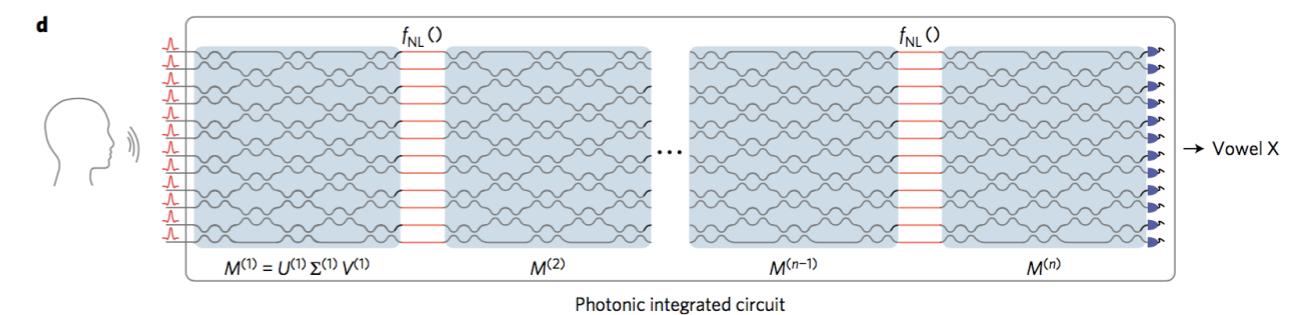
- ~mm = atomic quantum computers
- ~m = quantum clusters
- ~km = quantum internet

2) Integrated Photonics Foundries

- driven by classical comms.

3) Classical Applications

- low power computing
- signals processing
- imaging



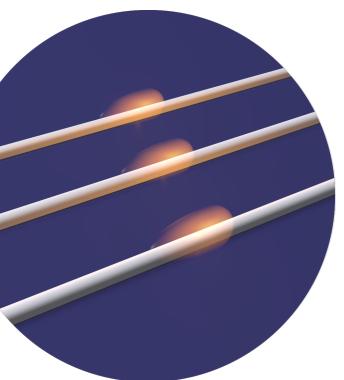
Y. Shen, N. Harris et al., *Nat. Photon.* (2017)

Overview

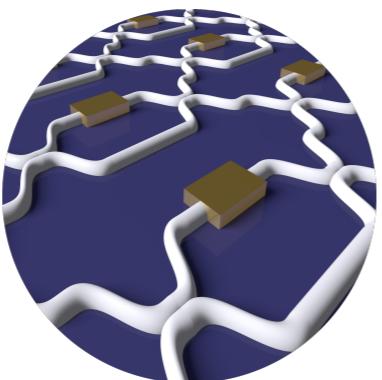
3. Quantum & AI

2. Systems Engineering Challenges

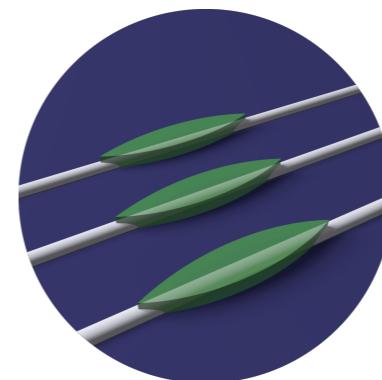
Applications



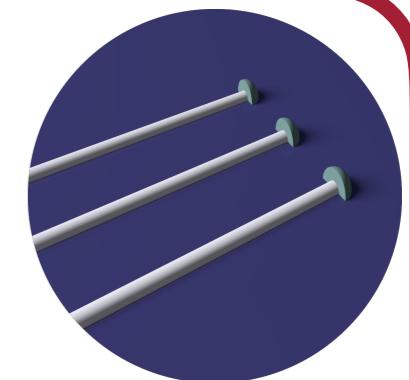
Nonclassical
states of light



Reconfigurable
optical circuitry

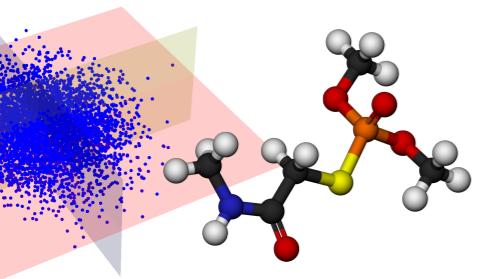


Nonlinear
light-matter interaction

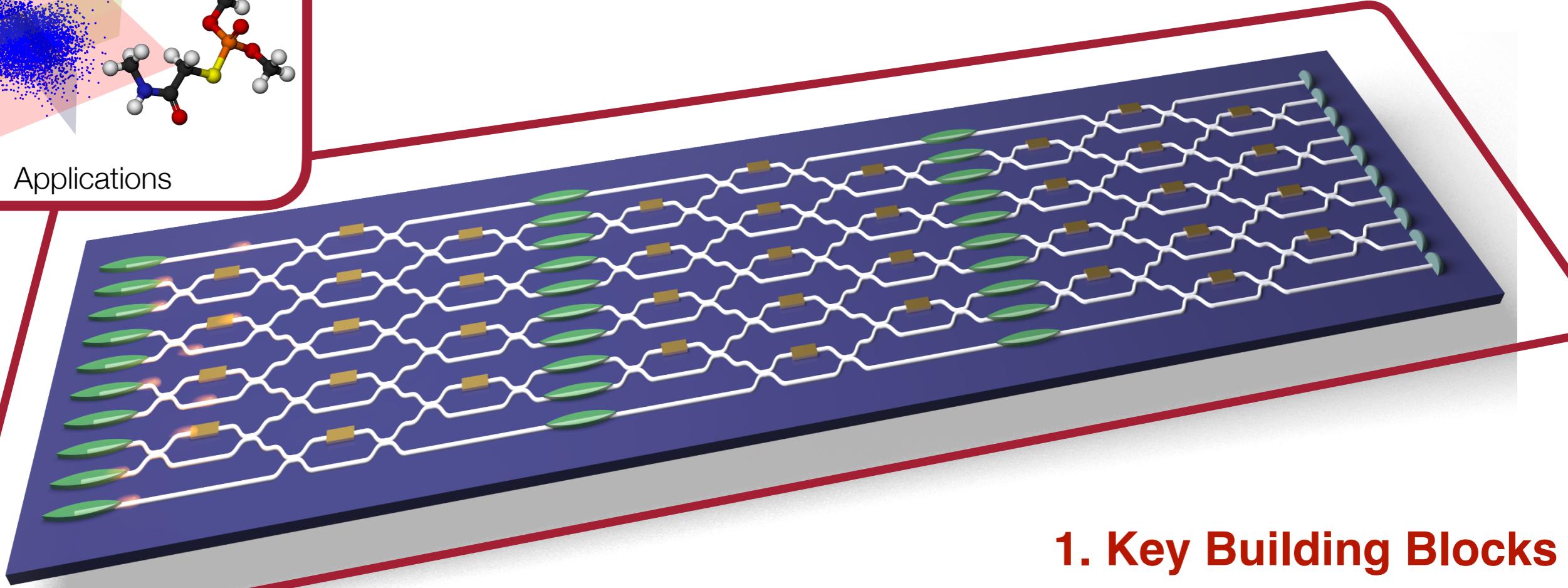


Single photon
readout

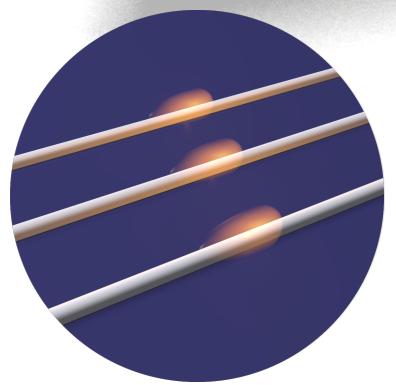
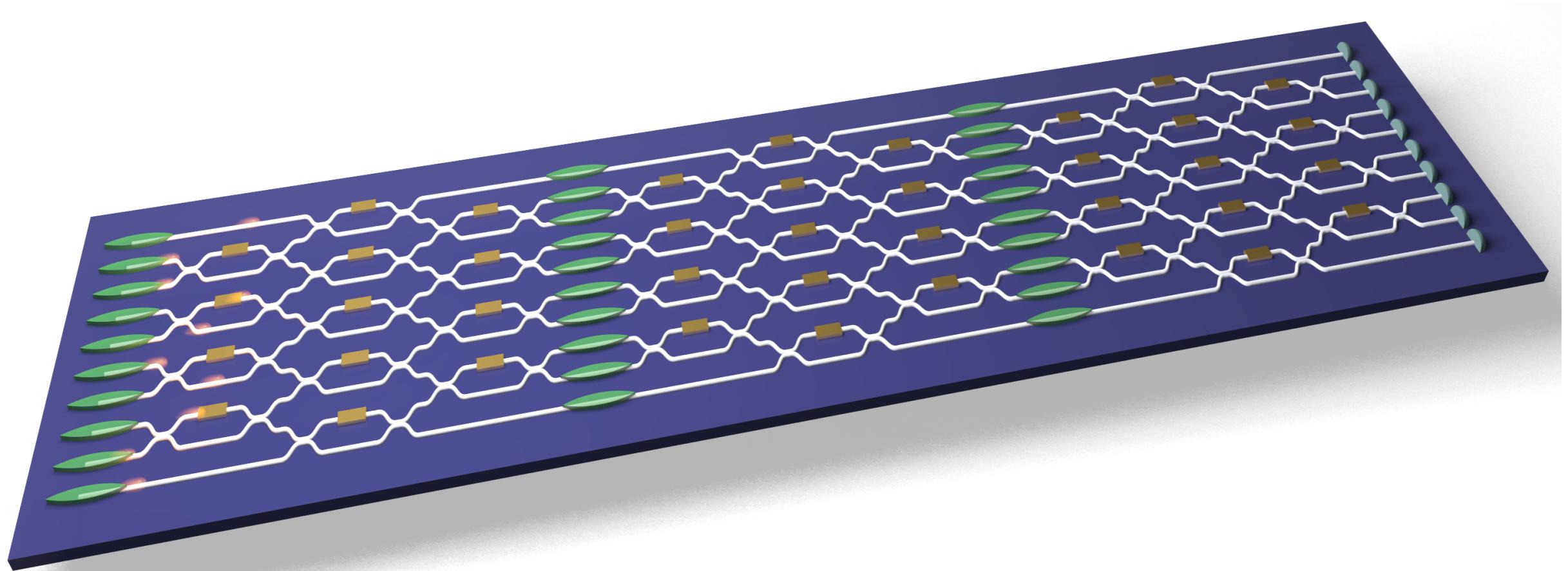
1. Key Building Blocks



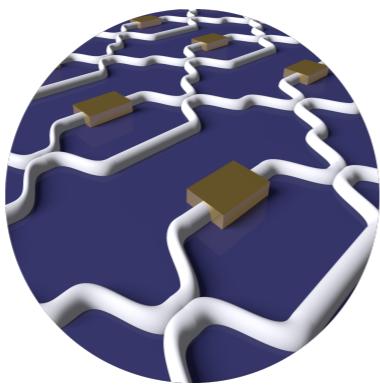
Applications



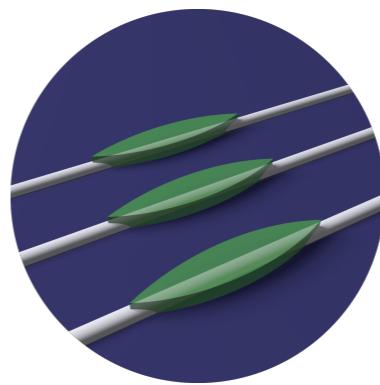
Quantum Photonic Processor



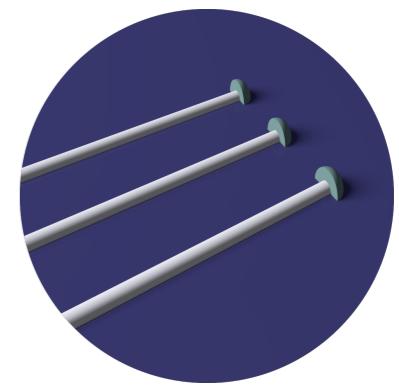
Nonclassical
states of light



Reconfigurable
optical circuitry

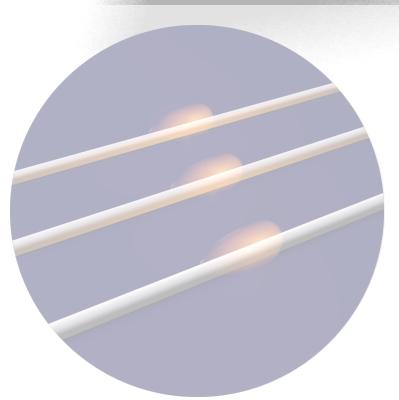
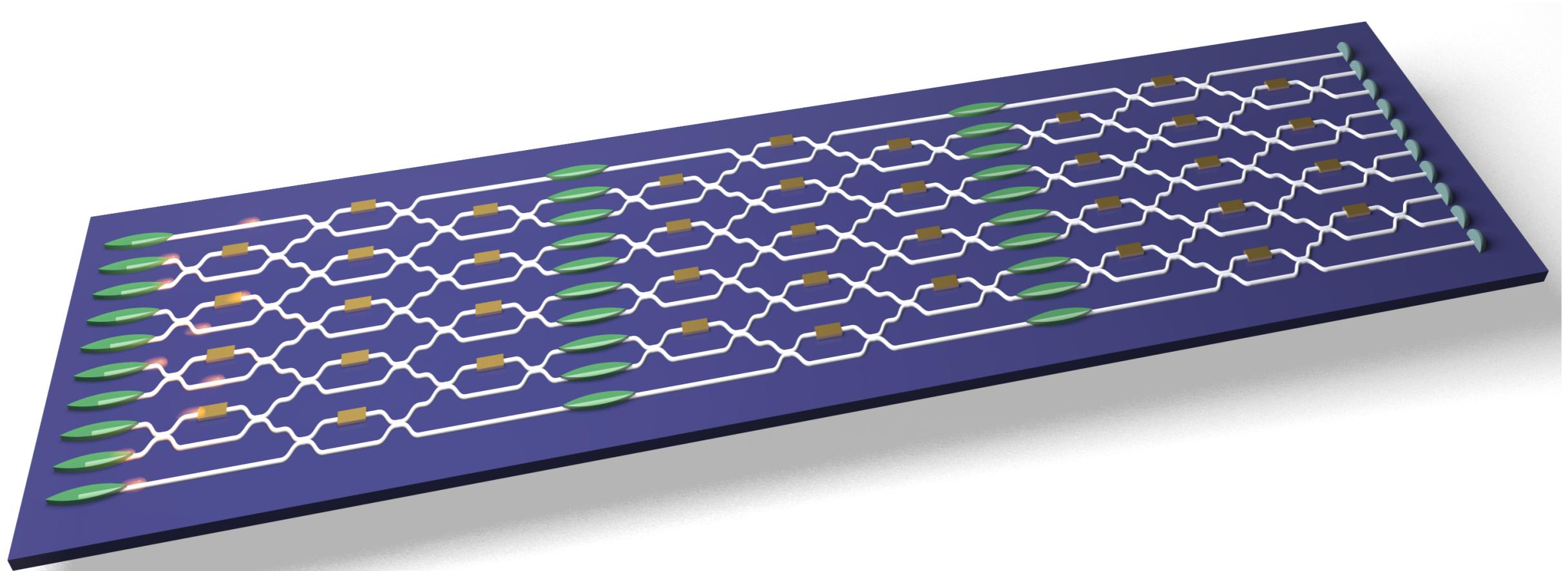


Nonlinear
light-matter interaction

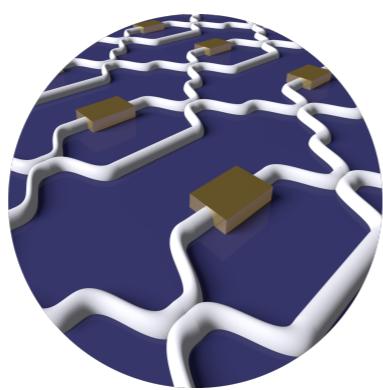


Single photon
readout

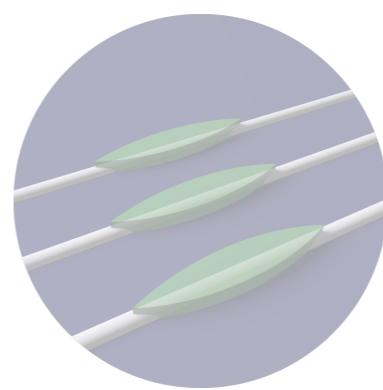
Quantum Photonic Processor



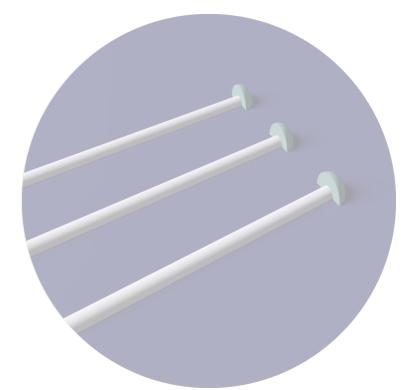
Nonclassical
states of light



Reconfigurable
optical circuitry



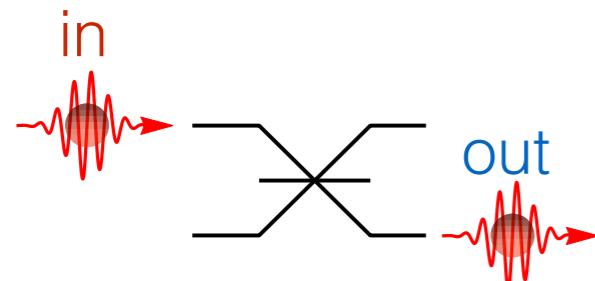
Nonlinear
light-matter interaction



Single photon
readout

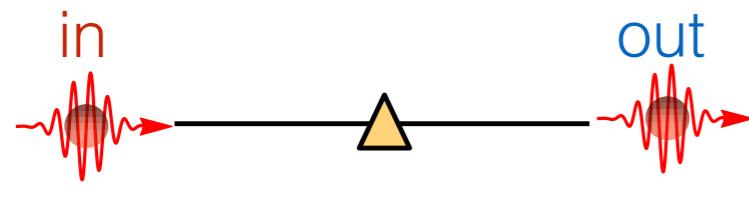
Building Blocks

Beamsplitters



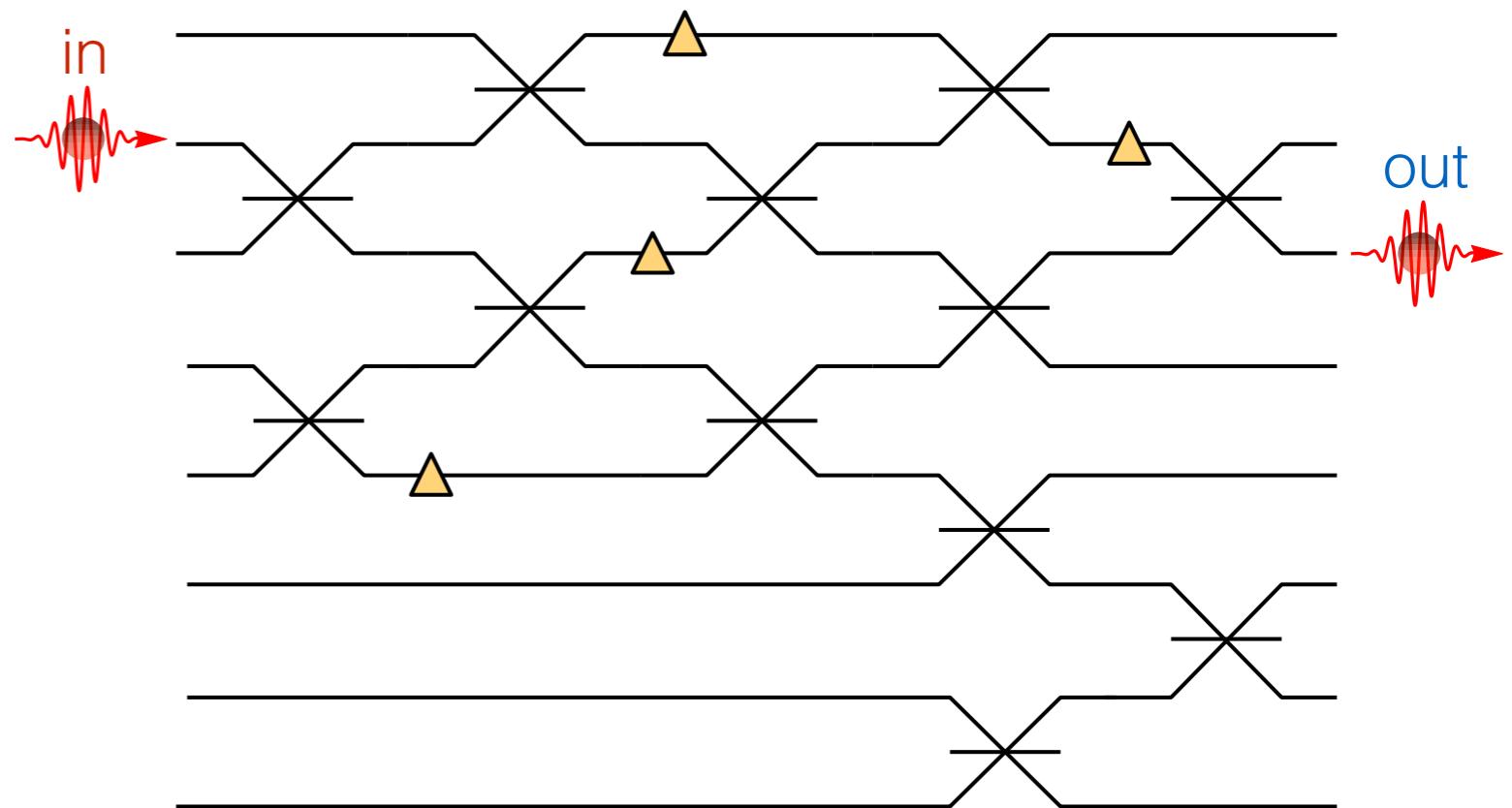
$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \text{out}$$

Phase shifters



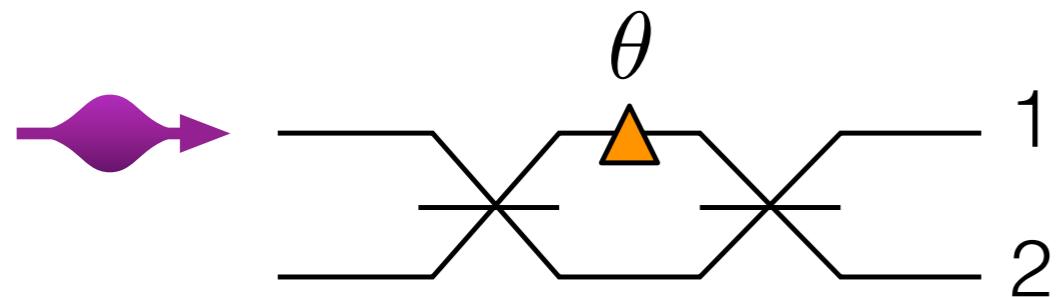
$$\begin{bmatrix} e^{i\phi} & 0 \\ 0 & 1 \end{bmatrix} \text{out}$$

Large circuits



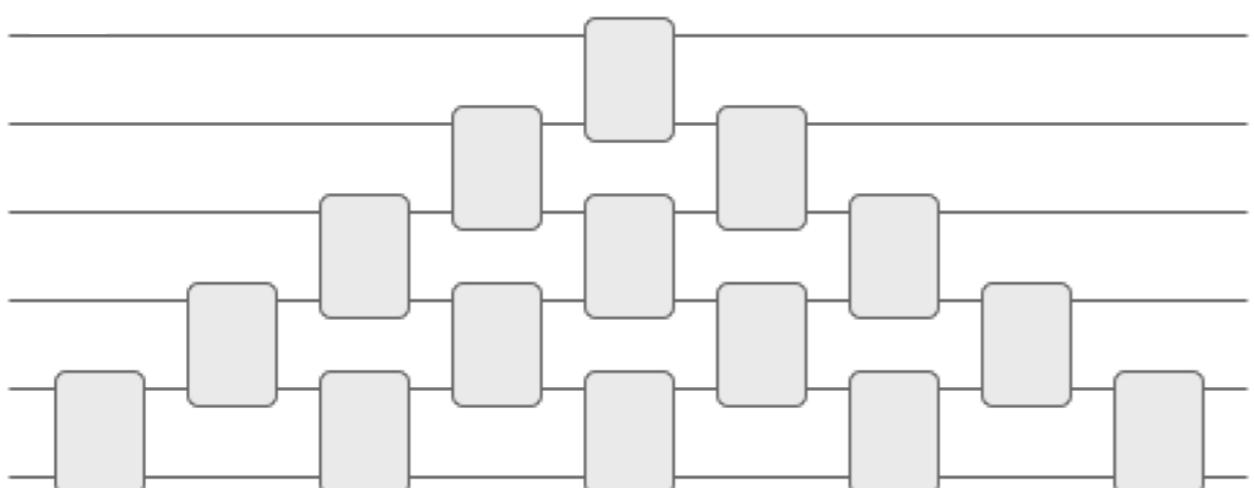
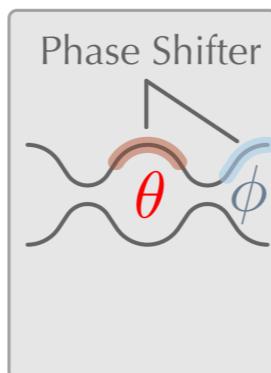
$$U = \begin{bmatrix} u_{1,1} & u_{1,2} & u_{1,2} & \cdots & u_{1,m} \\ u_{2,1} & u_{2,2} & u_{1,2} & \cdots & u_{2,m} \\ u_{3,1} & u_{3,2} & u_{1,2} & \cdots & u_{3,m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ u_{m,1} & u_{m,2} & u_{m,2} & \cdots & u_{m,m} \end{bmatrix} \begin{matrix} \text{in} \\ \text{out} \end{matrix}$$

Programmable Photonics

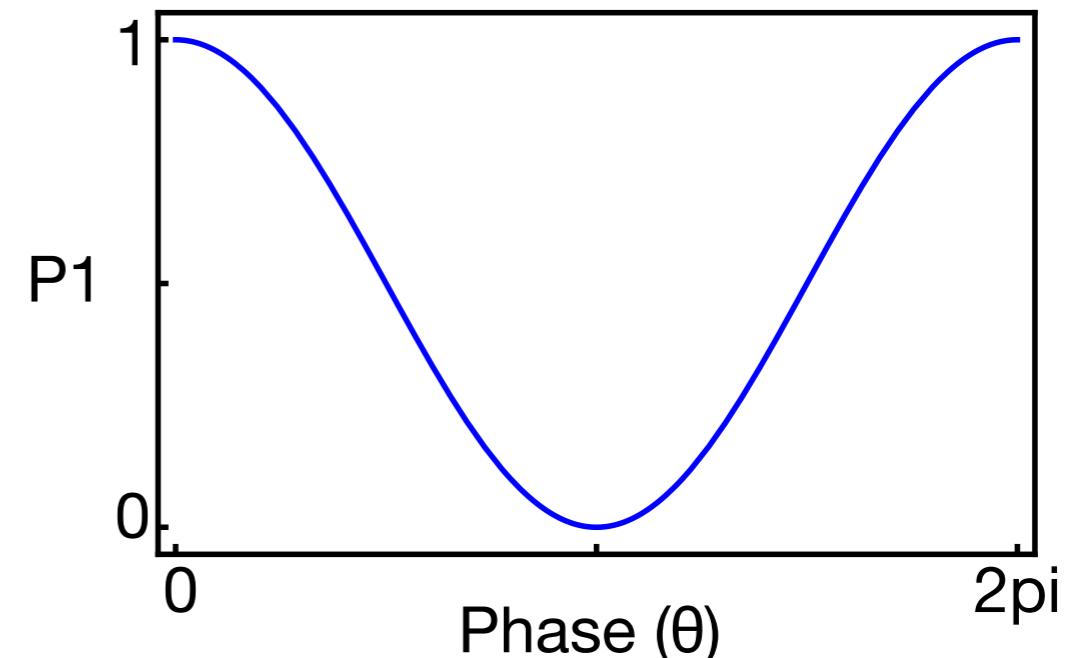


Mach-Zehnder Interferometer

Arbitrary $N \times N$ unitary
can be constructed out
of N^2 SU(2) MZIs



M. Reck et al., *PRL* (1994)



Review Article Vol. 5, No. 12 / December 2018 / *Optica* 1623

optica

Linear programmable nanophotonic processors

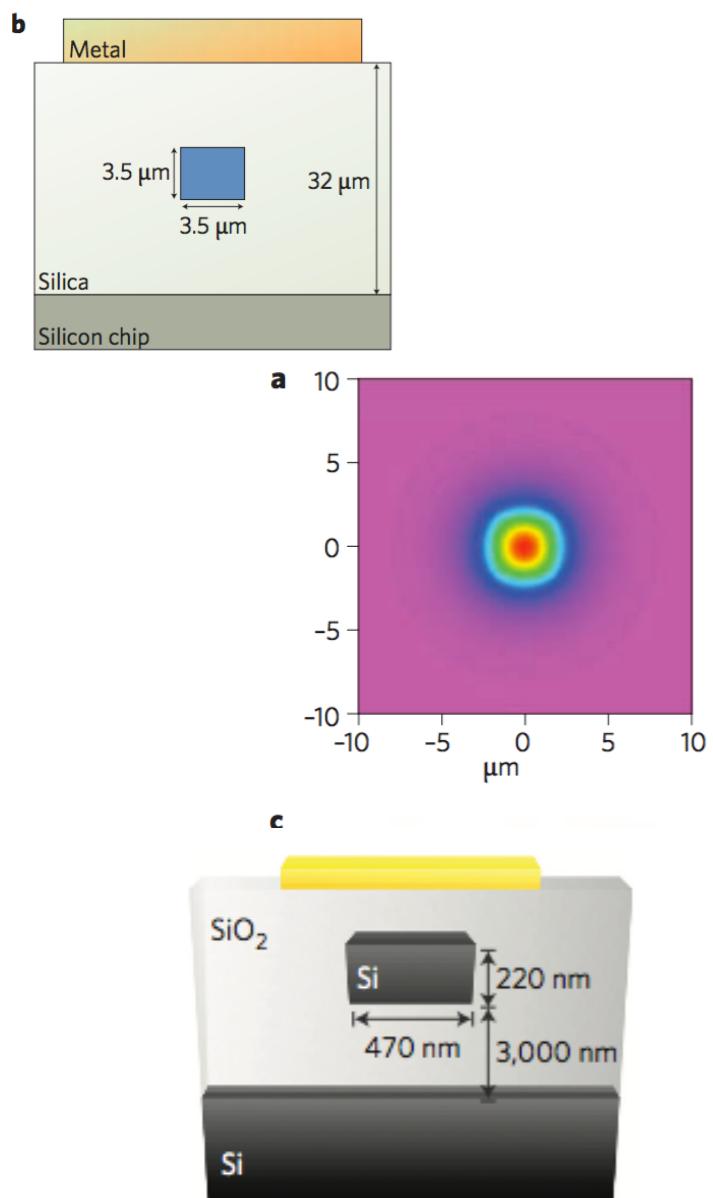
NICHOLAS C. HARRIS,¹ JACQUES CAROLAN,² DARIUS BUNANDAR,² MIHIKA PRABHU,² MICHAEL HOCHBERG,³ TOM BAEHR-JONES,³ MICHAEL L. FANTO,⁴ A. MATTHEW SMITH,⁴ CHRISTOPHER C. TISON,⁴ PAUL M. ALSING,⁴ AND DIRK ENGLUND^{2,*}

Integrated Photonics

Waveguide devices

Si, SiN, SiO₂, LN, AlN

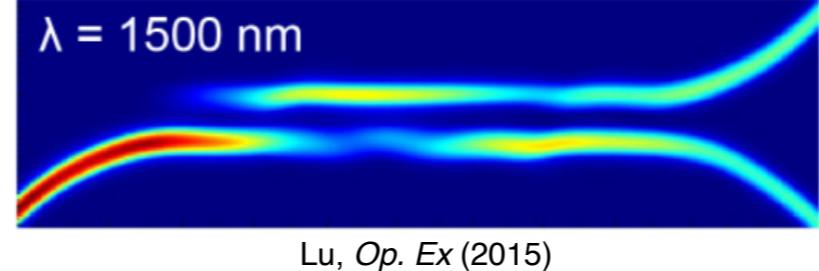
A. Politi et al., *Science*, 320, 646 (2008).
A. Politi et al., *Science*, 325, 1221 (2009).
J. W. Silverstone et al., *Nat. Photon.* 8, 104 (2014).



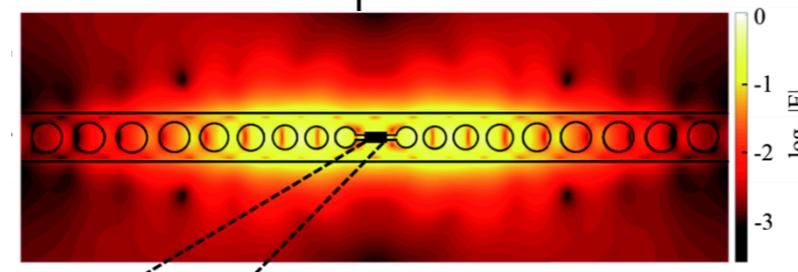
Passive Control

WG crossings
Grating couplers
Edge couples
Microring resonators

Directional Couplers

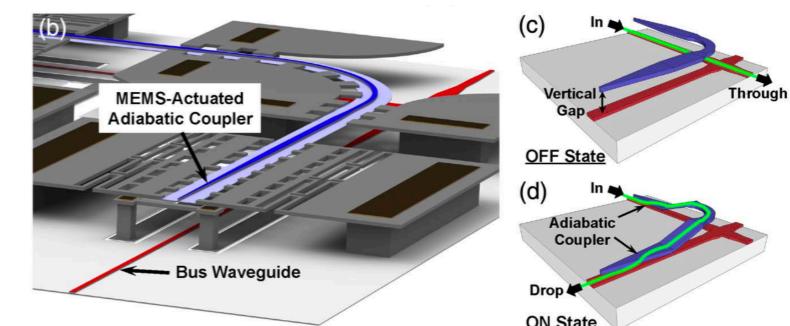
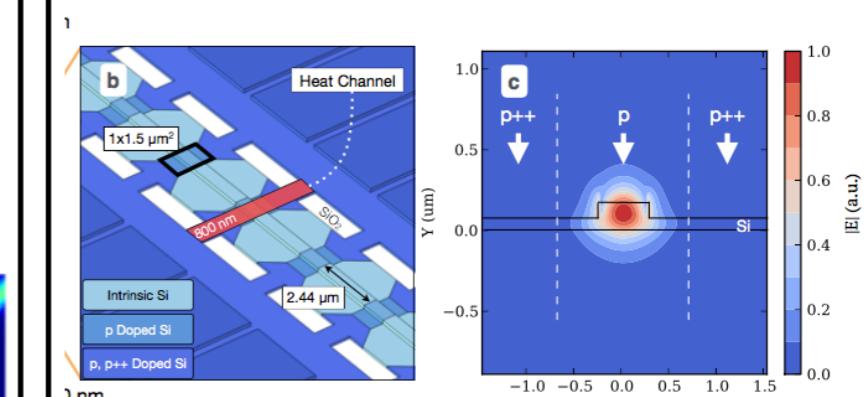


On-chip Cavities



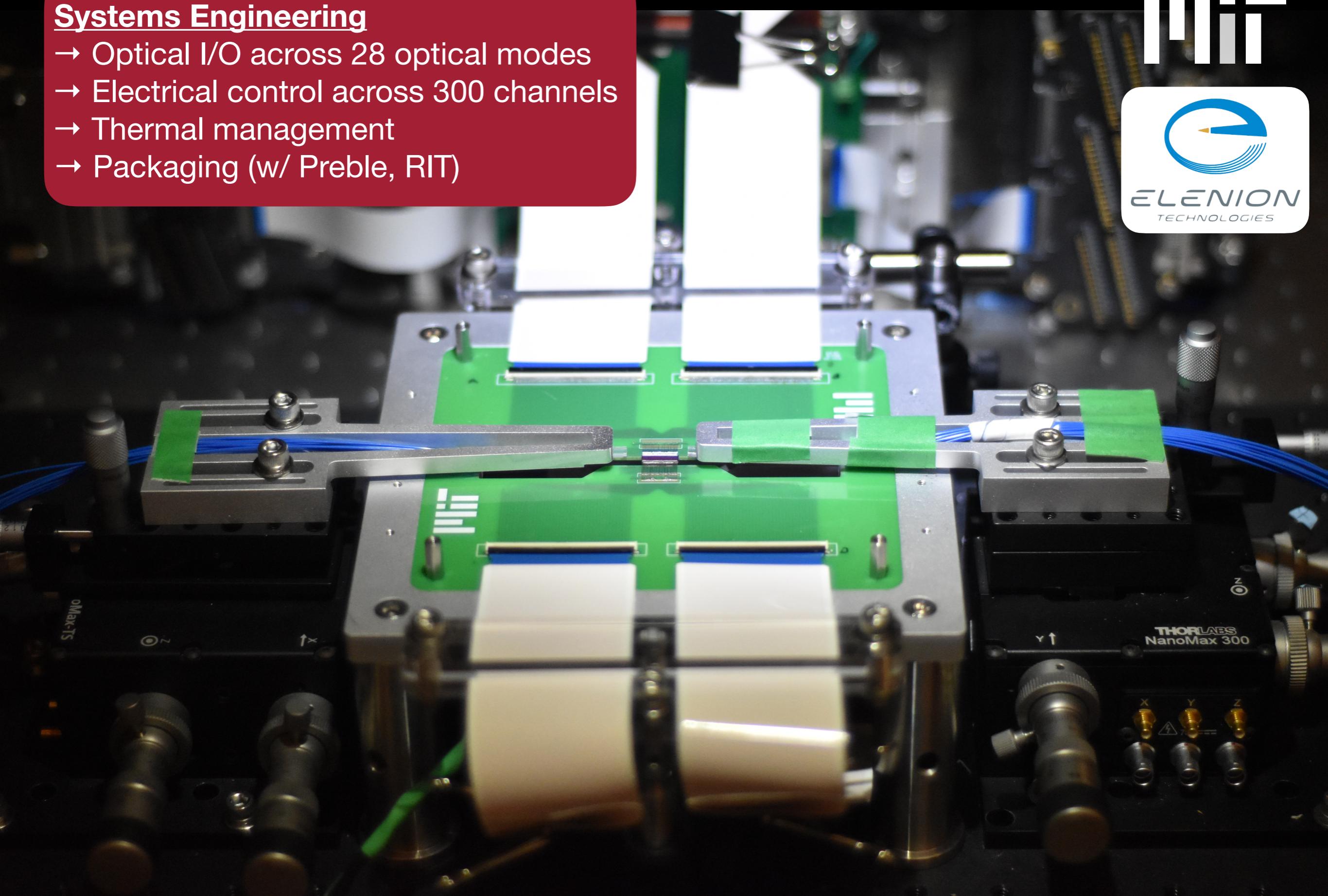
Active control

Thermal (KHz)
MEMS (MHz)
Electro-optic (GHz)

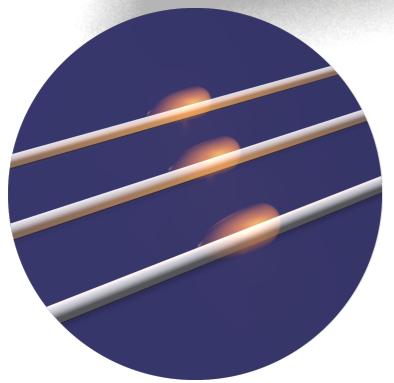
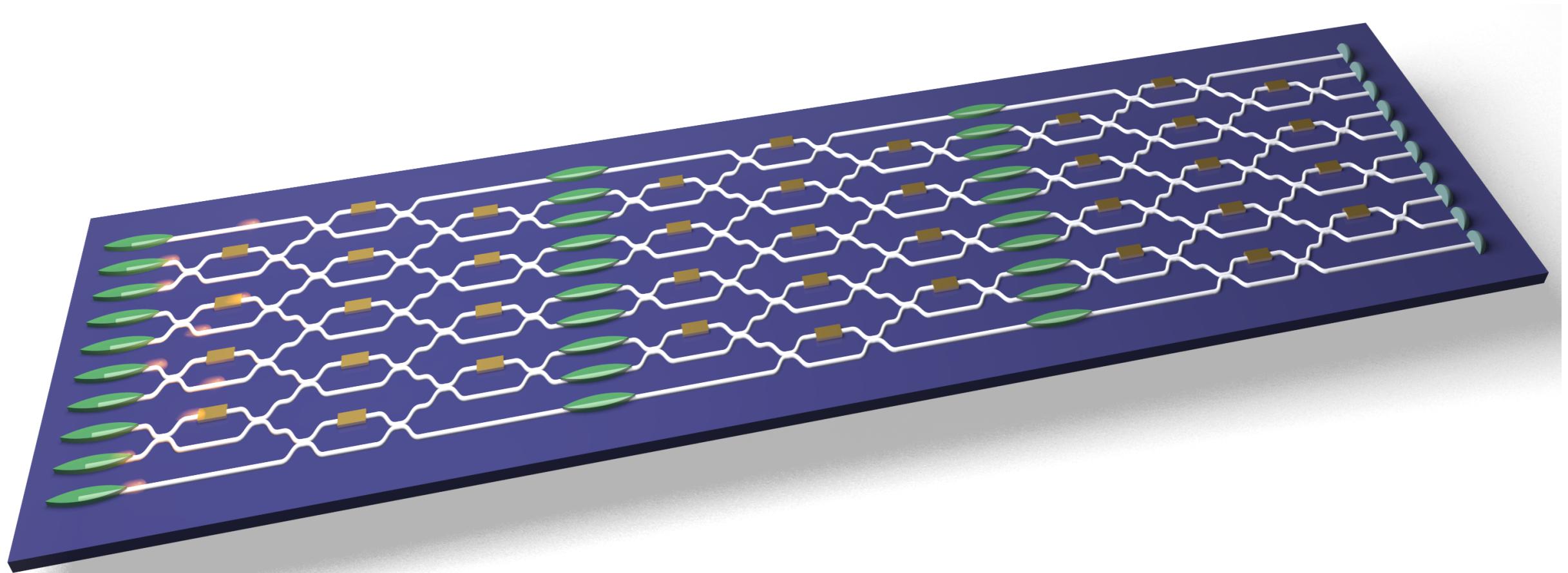


Systems Engineering

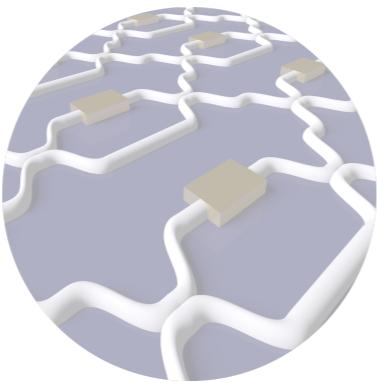
- Optical I/O across 28 optical modes
- Electrical control across 300 channels
- Thermal management
- Packaging (w/ Preble, RIT)



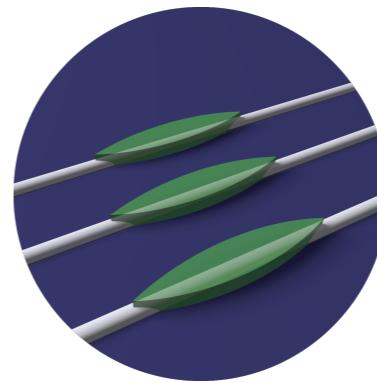
Quantum Photonic Processor



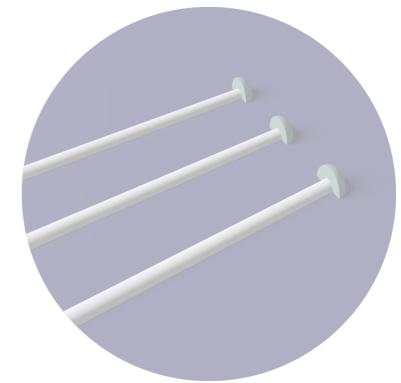
Nonclassical
states of light



Reconfigurable
optical circuitry



Nonlinear
light-matter interaction



Single photon
readout

III/V Semiconductor QDs

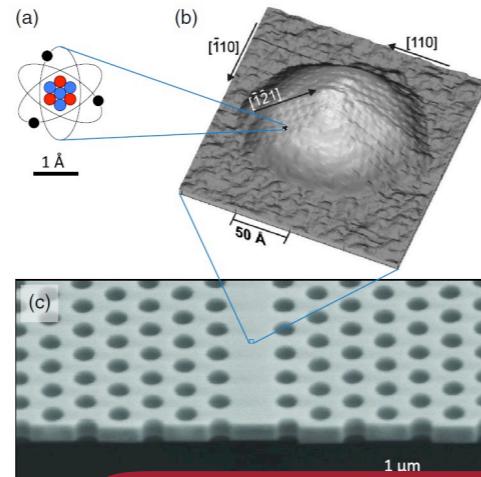


P. Lodahl, NBI

Artificial Atoms



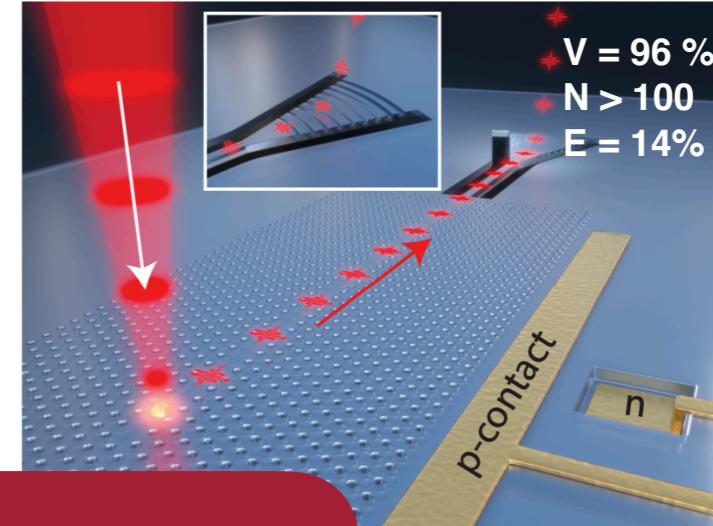
KØBENHAVNS
UNIVERSITET



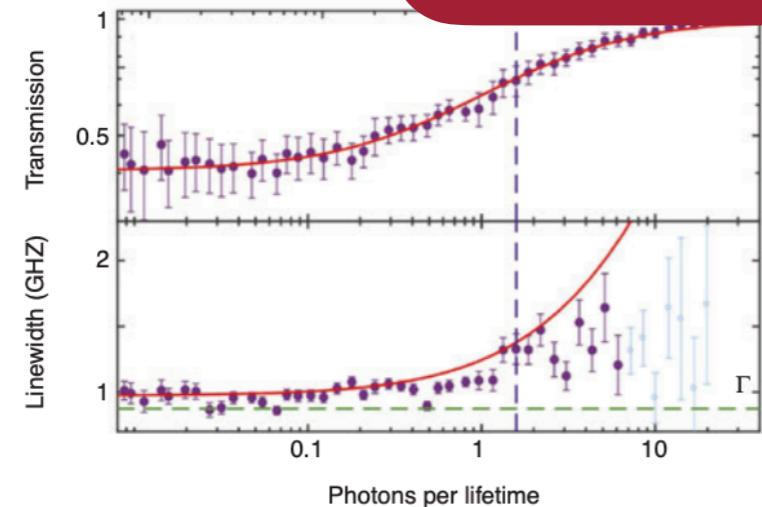
P. Lodahl et al., Rev.

- Spatial variation
- Spectral variation
- Compatibility with foundry PICs

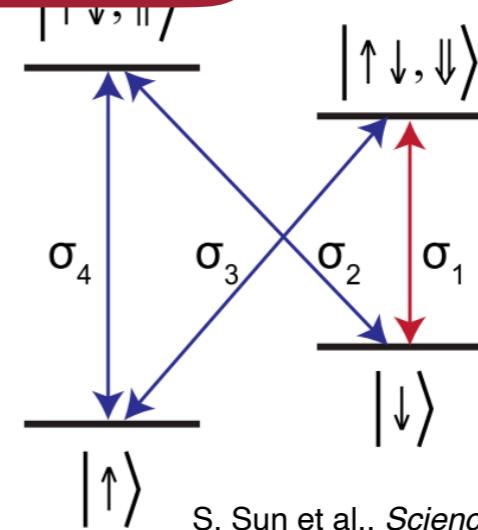
Efficient Single Photon Sources



Single Photon Switch

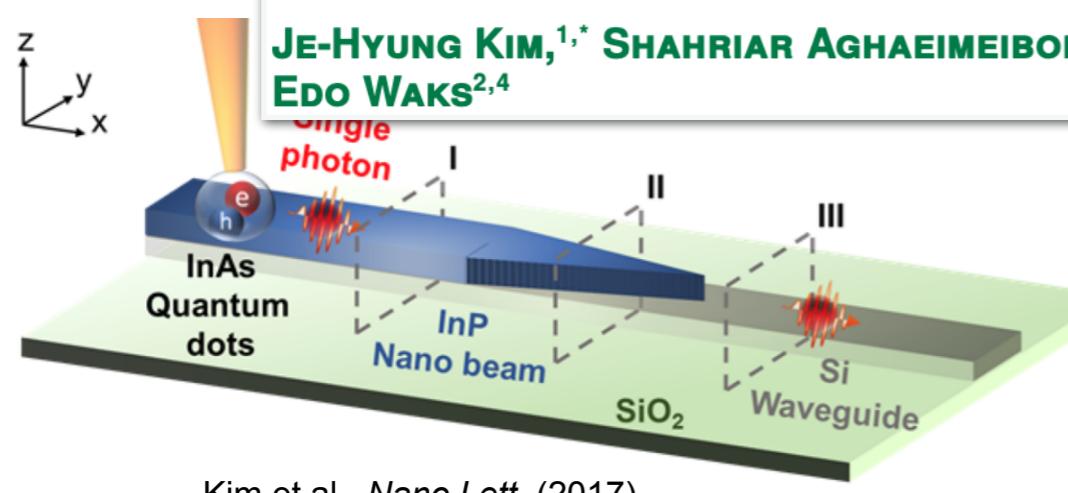


H. Thyrrstrup et al., *Nano. Lett.* (2018)



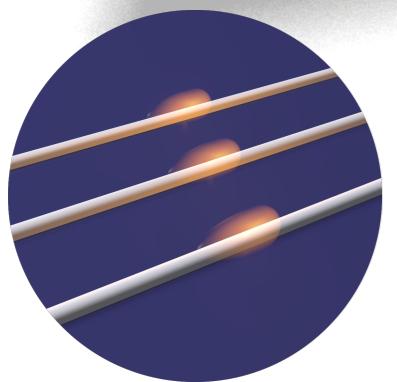
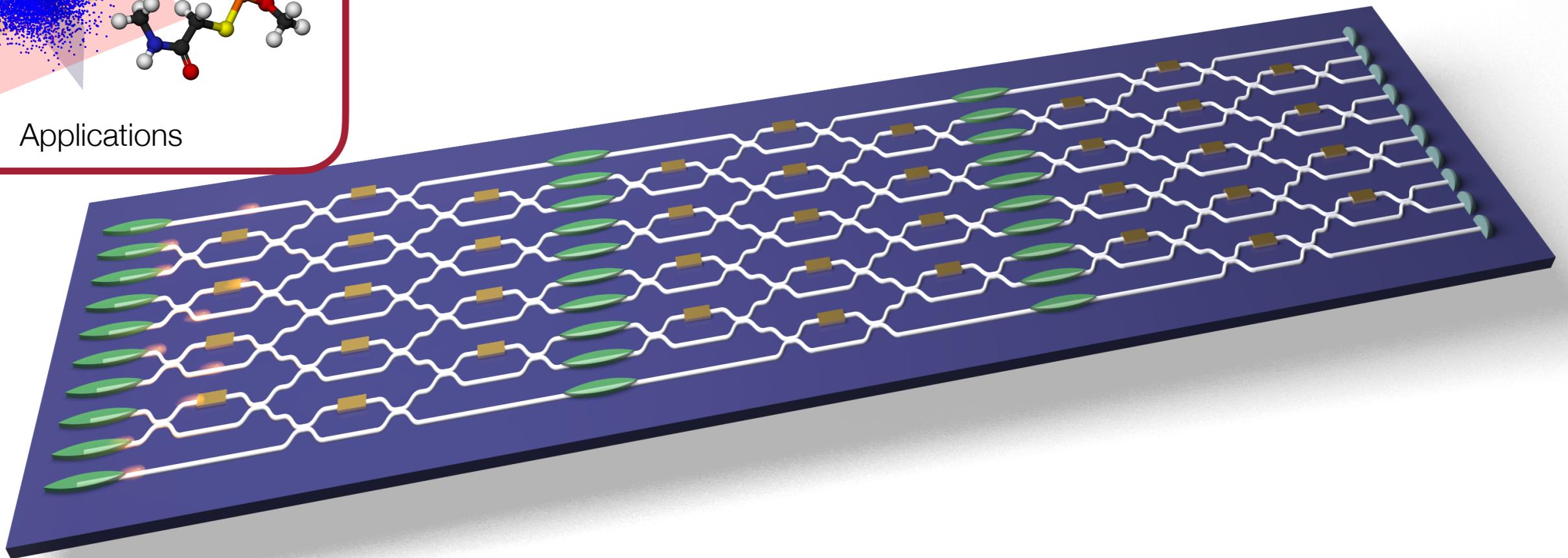
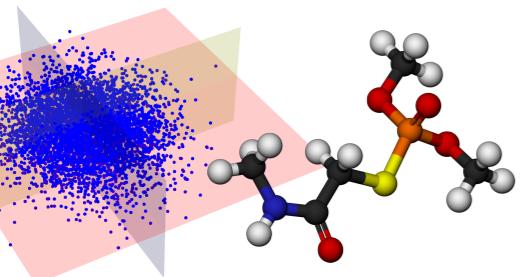
S. Sun et al., *Science* (2018)

Hybrid Integration

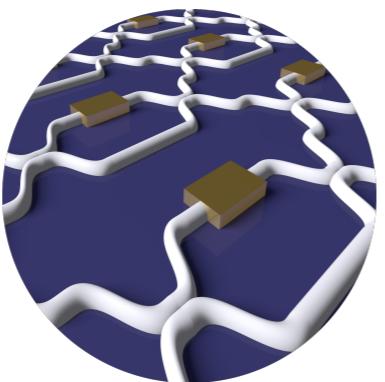


- Pre-select QDs (frequency, bandwidth)
- Use PIC for control (strain, Stark)
- Wafer-wafer bonding

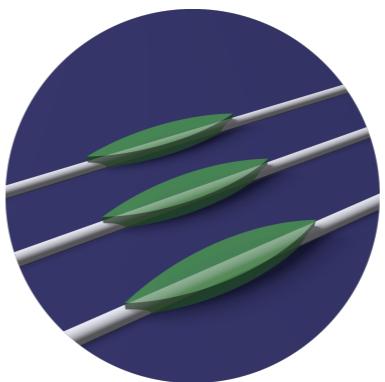
Quantum Photonic Processor



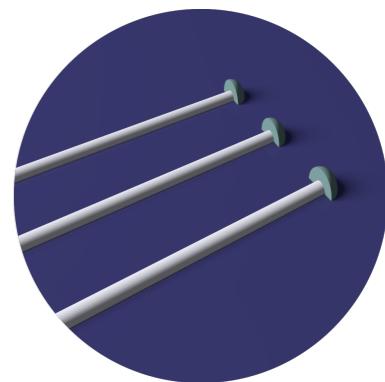
Nonclassical
states of light



Reconfigurable
optical circuitry



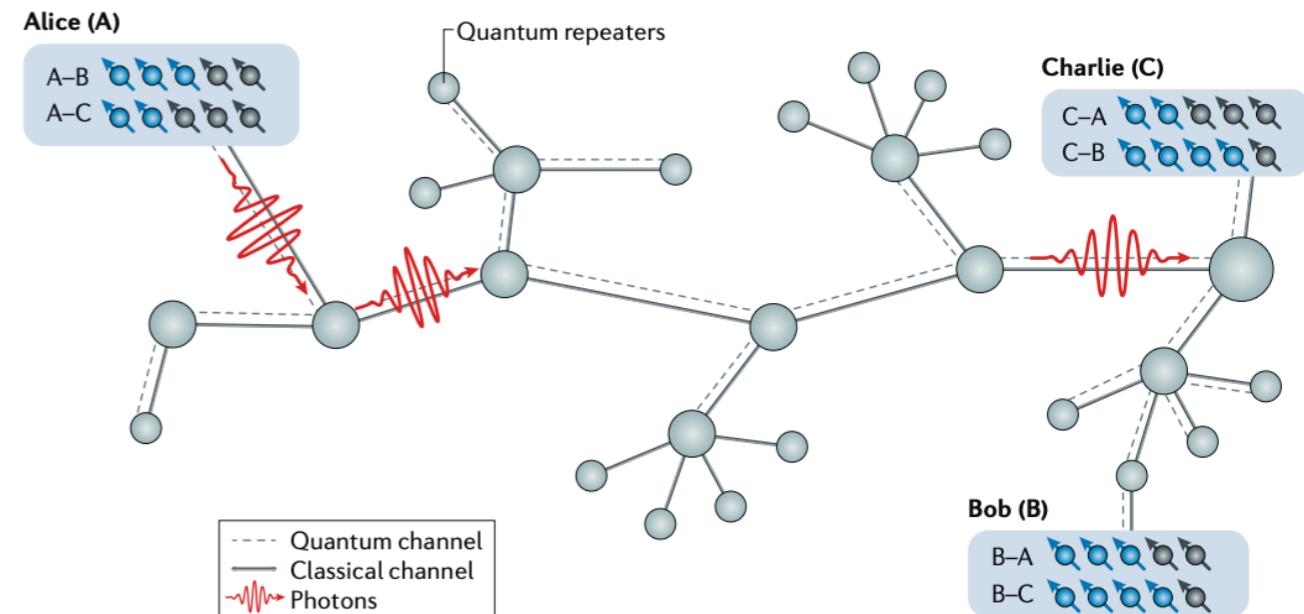
Nonlinear
light-matter interaction



Single photon
readout

Loss Correction

- Primary error mechanism is **loss**
- Encode a single qubit of information across multiple photons s.t.
if the error occurs information can still be recovered!
- **One-way quantum repeaters** do this without measurement or quantum memories



Codeword¹

$$|0\rangle_L \equiv (|40\rangle_{12} + |04\rangle_{12}) / \sqrt{2}$$
$$|1\rangle_L \equiv |22\rangle_{12},$$

System

$$\hat{S} |30\rangle_{12} = (|40\rangle_{12} + |04\rangle_{12}) / \sqrt{2}$$
$$\hat{S} |03\rangle_{12} = (|40\rangle_{12} + |04\rangle_{12}) / \sqrt{2}$$
$$\hat{S} |12\rangle_{12} = |22\rangle_{12}$$
$$\hat{S} |21\rangle_{12} = |22\rangle_{12}.$$

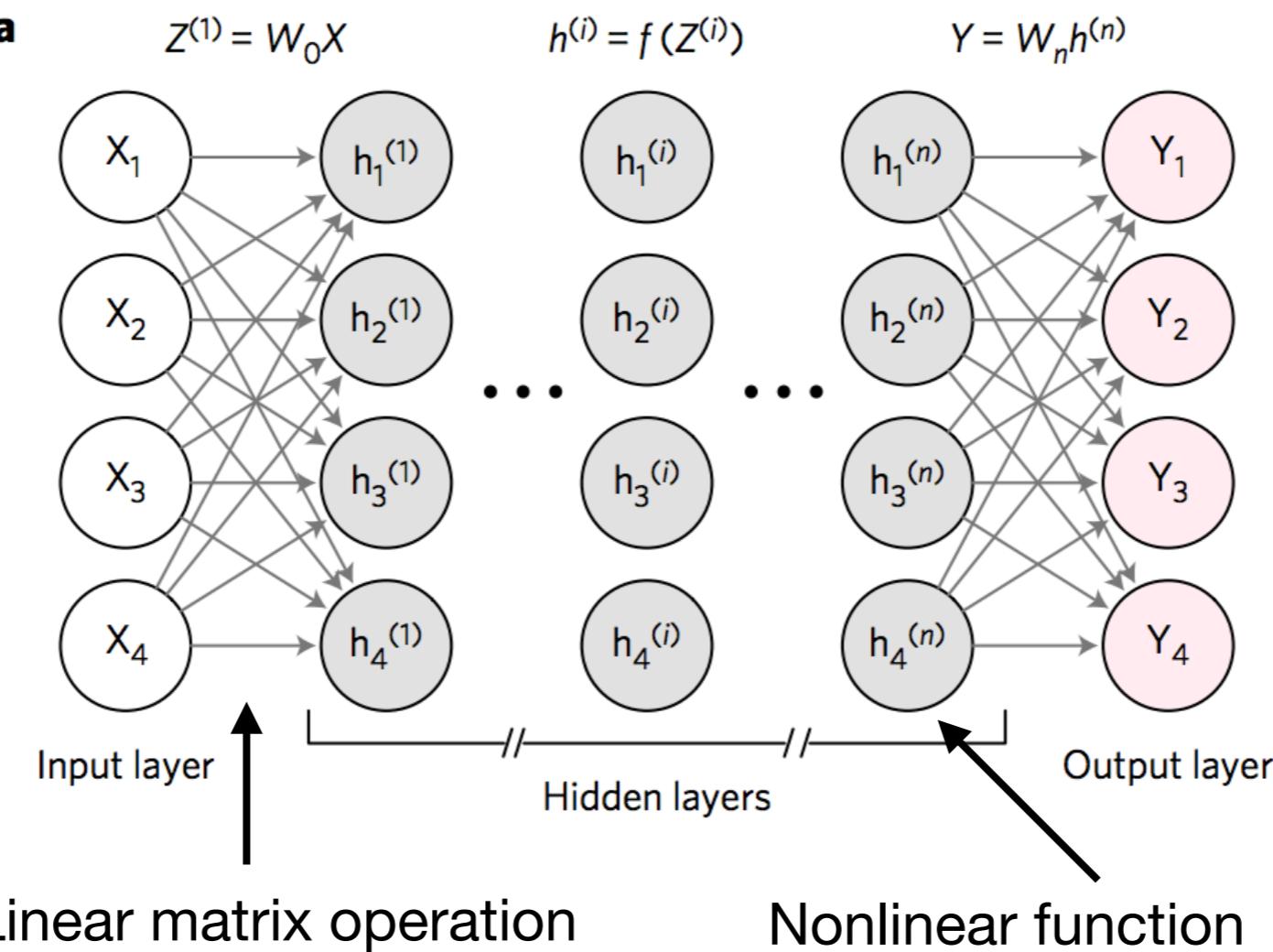
“we do not yet understand how to determine if a given Hamiltonian can be implemented (or approximated to a satisfactory degree) by arranging a reasonable number of optical components”

F. M. Miatto et al., Quantum 2, 75 (2018)

Neural Networks

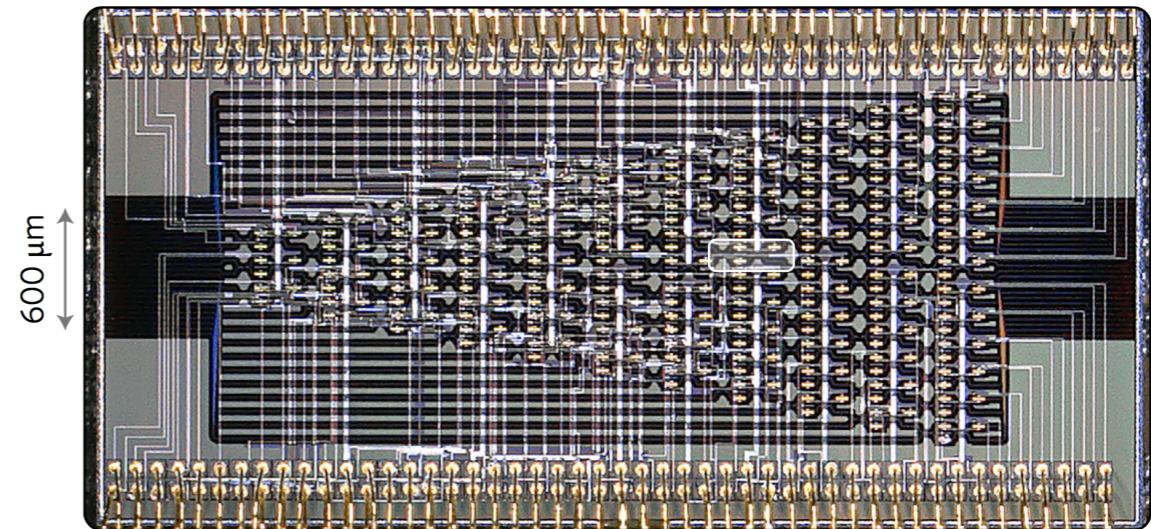


$|30\rangle$



Optical Neural Networks

- Matrix Multiplication
- High bandwidth (100 GHz photodiodes)
- Low power (quasi-static phase control)



nature
photonics

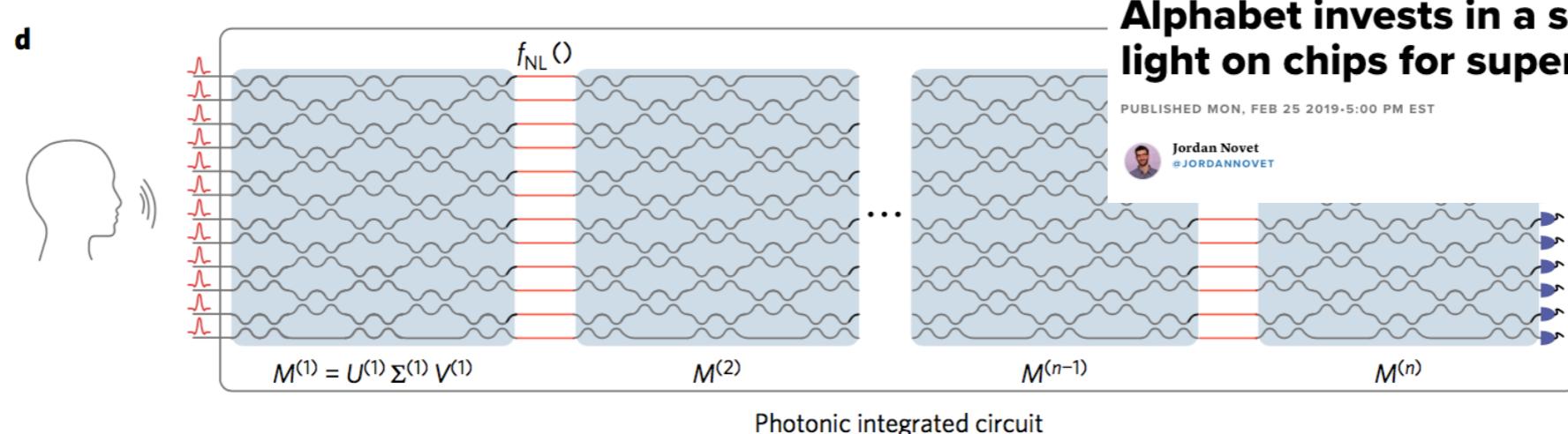
ARTICLES

PUBLISHED ONLINE: 12 JUNE 2017 | DOI: 10.1038/NPHOTON.2017.93

15.
2017).
(2018).

Deep learning with coherent nanophotonic circuits

Yichen Shen^{1*}, Nicholas C. Harris^{1*}, Scott Skirlo¹, Mihika Prabhu¹, Tom Baehr-Jones², Michael Hochberg², Xin Sun³, Shijie Zhao⁴, Hugo Larochelle⁵, Dirk Englund¹ and Marin Soljačić¹



Alphabet invests in a start-up using beams of light on chips for super-fast A.I.

PUBLISHED MON, FEB 25 2019 5:00 PM EST



Jordan Novet
@JORDANNOVET

SHARE f t in e



Quantum Optical Neural Networks

ARTICLE

OPEN

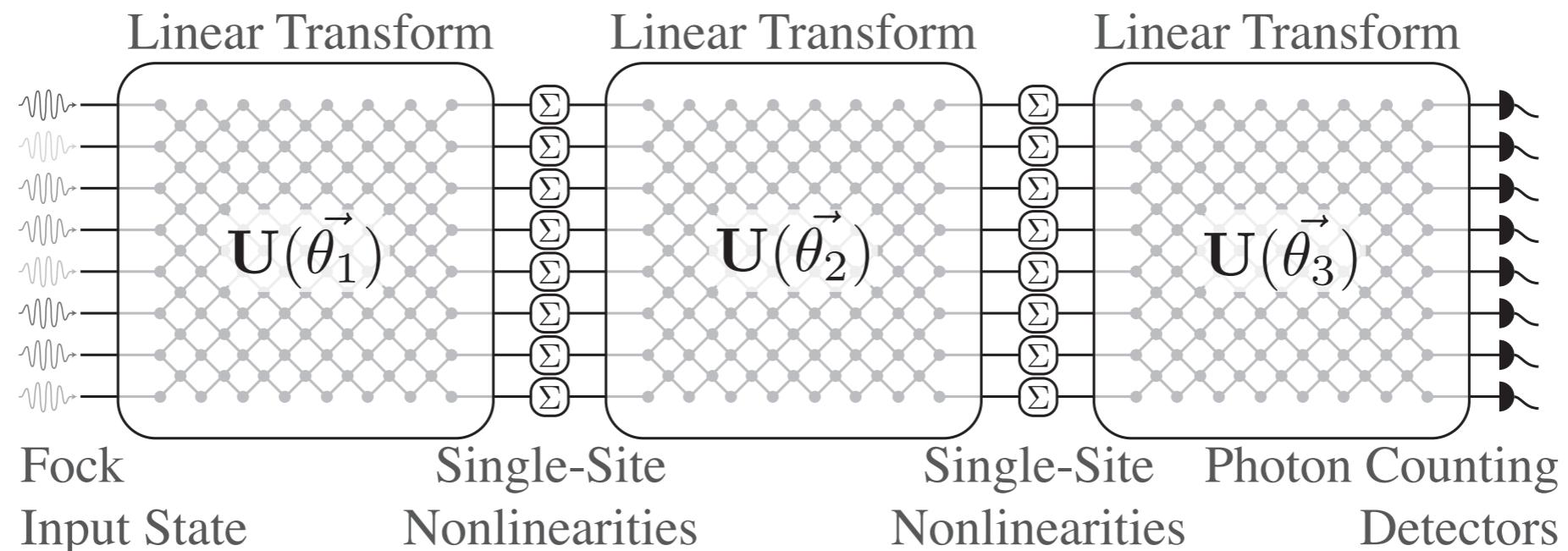
npj Quantum Information

Quantum optical neural networks

Gregory R. Steinbrecher¹, Jonathan P. Olson², Dirk Englund¹  and Jacques Carolan¹

(b) A Quantum Optical Neural Network

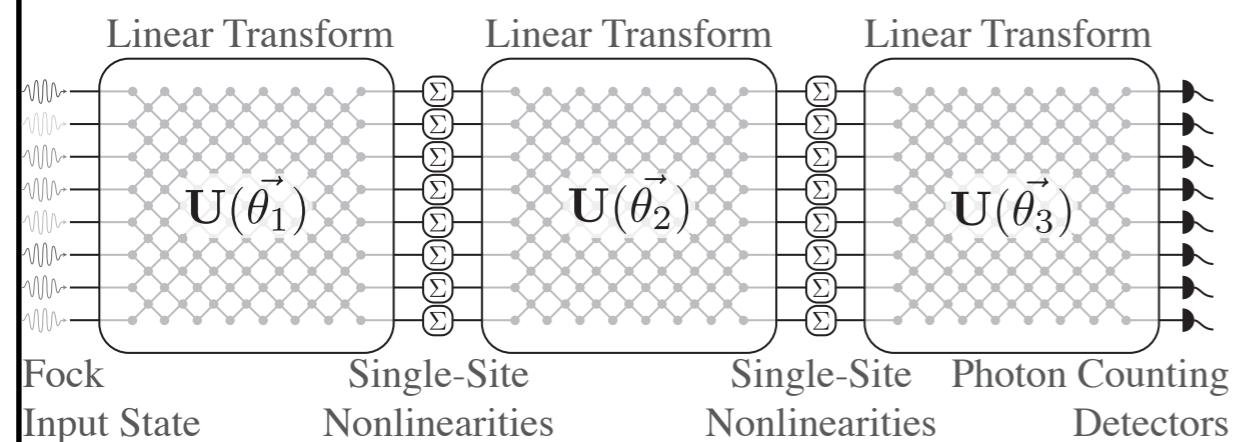
$$\Sigma(\phi) = \sum_{n=0}^{\infty} e^{in(n-1)\phi/2} |n\rangle \langle n|$$



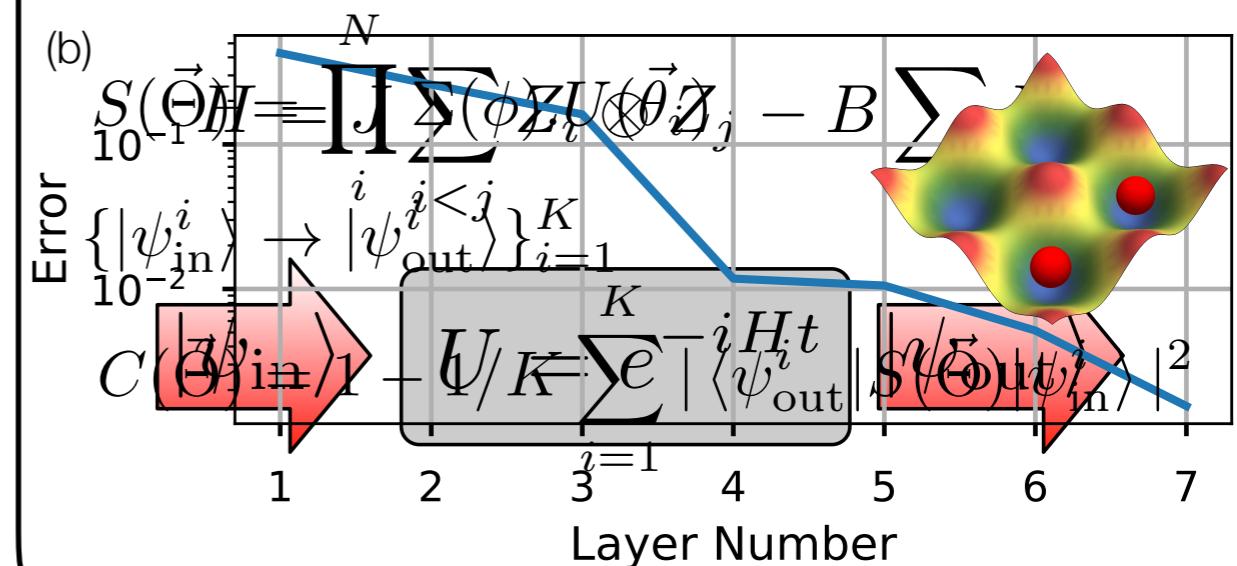
Greg Steinbrecher

Applications

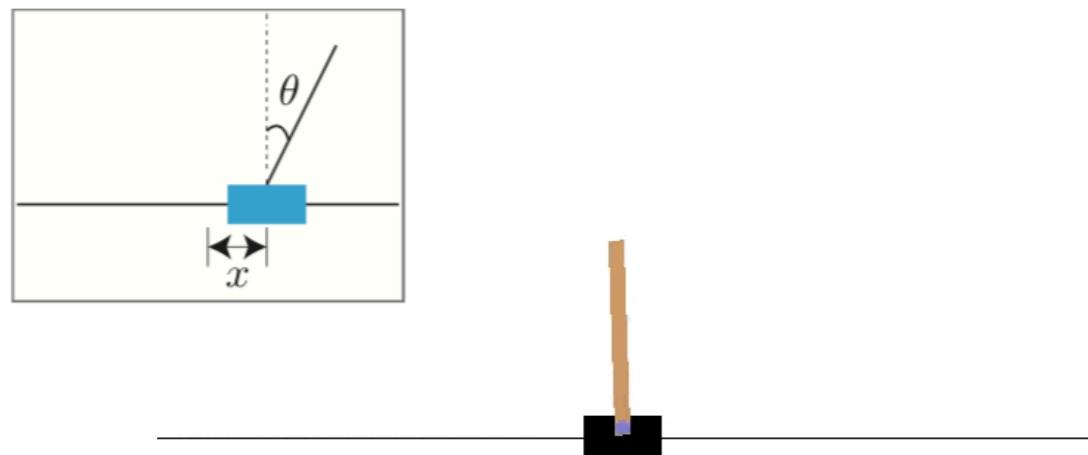
Training



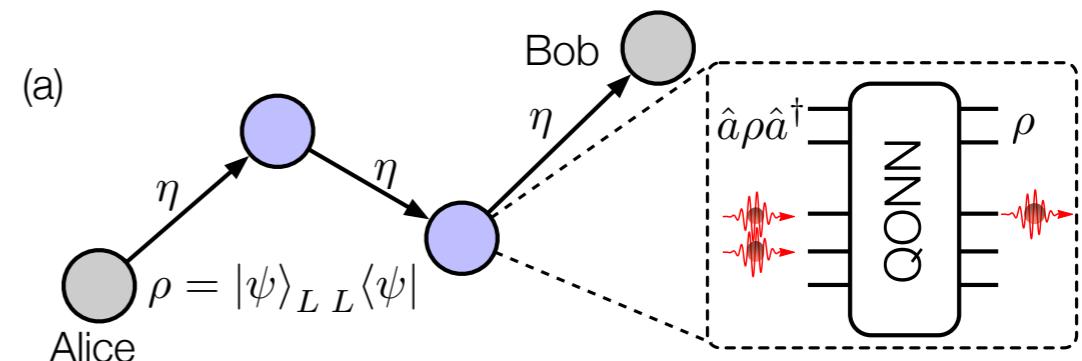
Black Box Quantum Simulation



Reinforcement Learning



One Way Quantum Repeater!



NISQ

- Noisy Intermediate Scale Quantum (NISQ) processors¹:

- Shallow depth
- Small [$O(100)$] qubits
- Error above threshold

- **Quantum Advantage:**

- Random Circuit Sampling
Bouland et al., Nat. Phys. (2018)
- IQP sampling
Bremner et al., Quantum (2016)
- Boson Sampling
Aaronson & Arkhipov (2014)

- **Near-term quantum algorithms:**

- QAOA (Max Cut)
Farhi et al., arXiv:1411.4028
- Variational Quantum Eigensolver
Peruzzo et al, Nat. Comms (2014)
- Quantum Autoencoder
Romero et al., QST (2017)

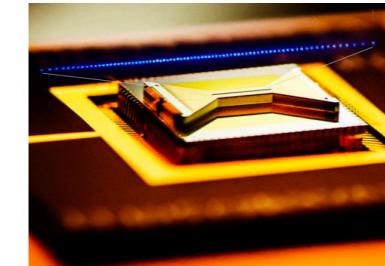
When you see the ratio of physical to logical qubits for fault tolerant quantum computation



[1] Quantum Computing Memes for QMA-complete Teens, Proceedings of the Royal Society of Quality Memes, Vol 5., 2018.



Google AI



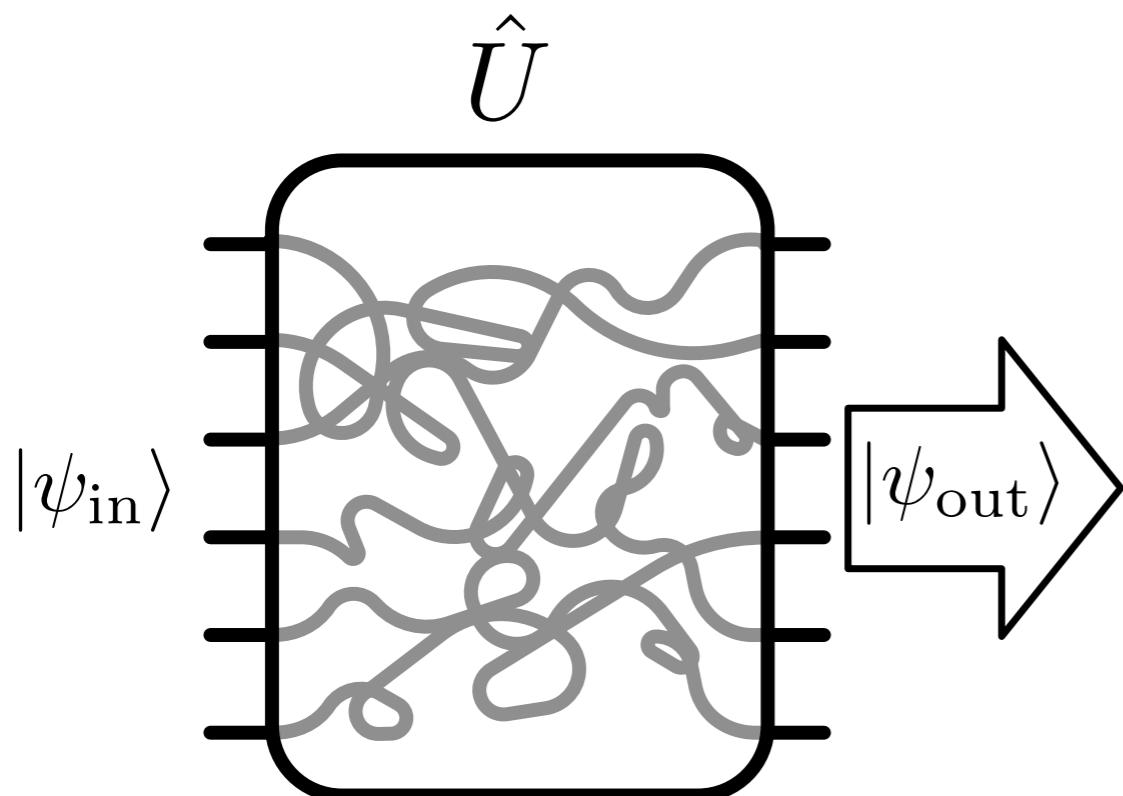
IONQ



IBM Q™

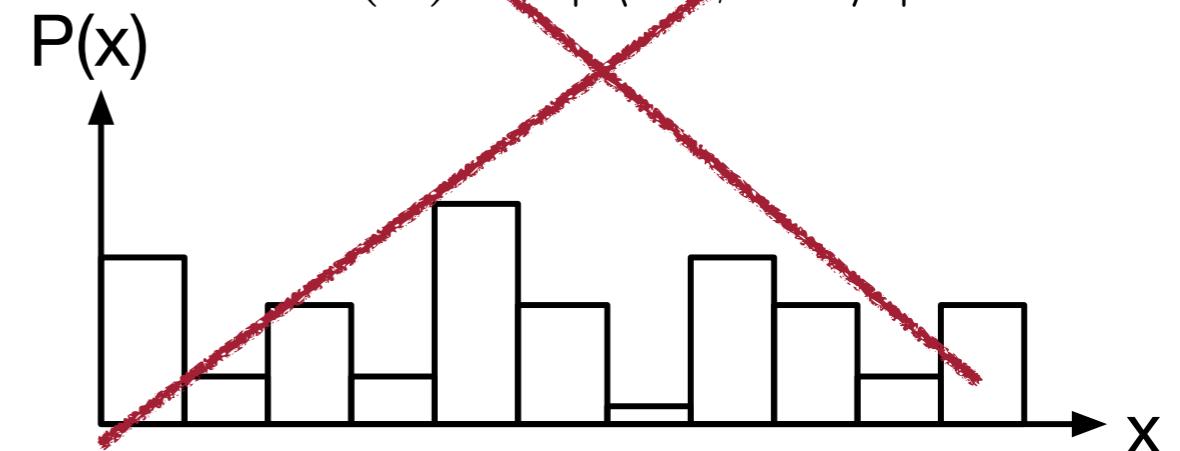
¹ J. Preskill, Quantum 2, 79 (2018)

Quantum Machine Learning



Given $\Psi_{in} + U$, sample from P_U

$$P_U(x) = |\langle x|\psi_{out}\rangle|^2$$



Given Ψ_{out} , what can we learn about U ?

nature
physics

ARTICLES

<https://doi.org/10.1038/s41567-019-0747-6>

Variational quantum unsampling on a quantum photonic processor

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Dr. Masoud Mohseni



Dr. Jonny Olson

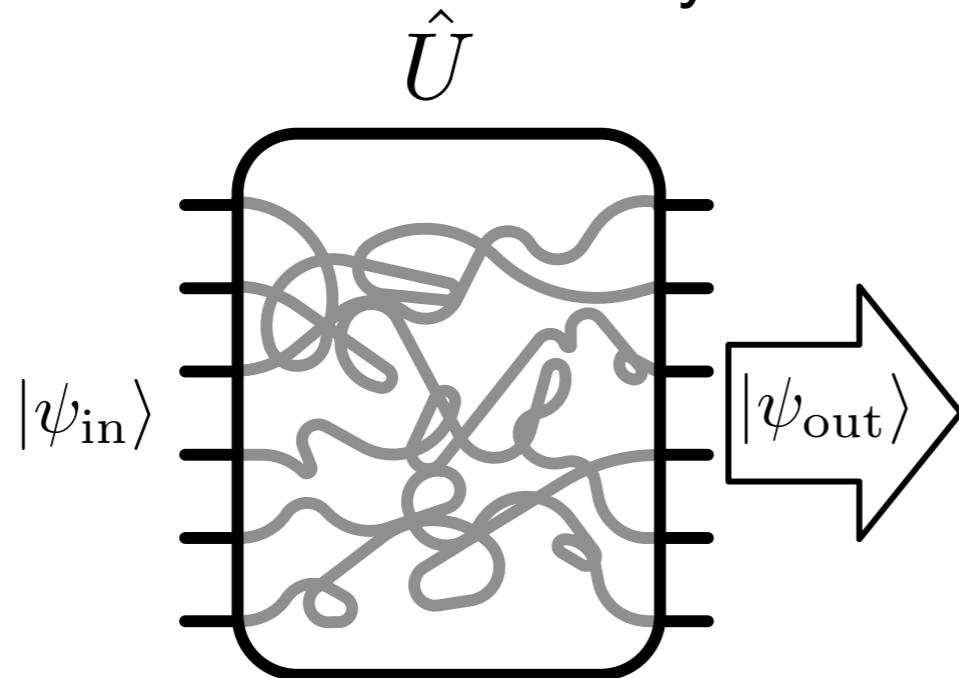


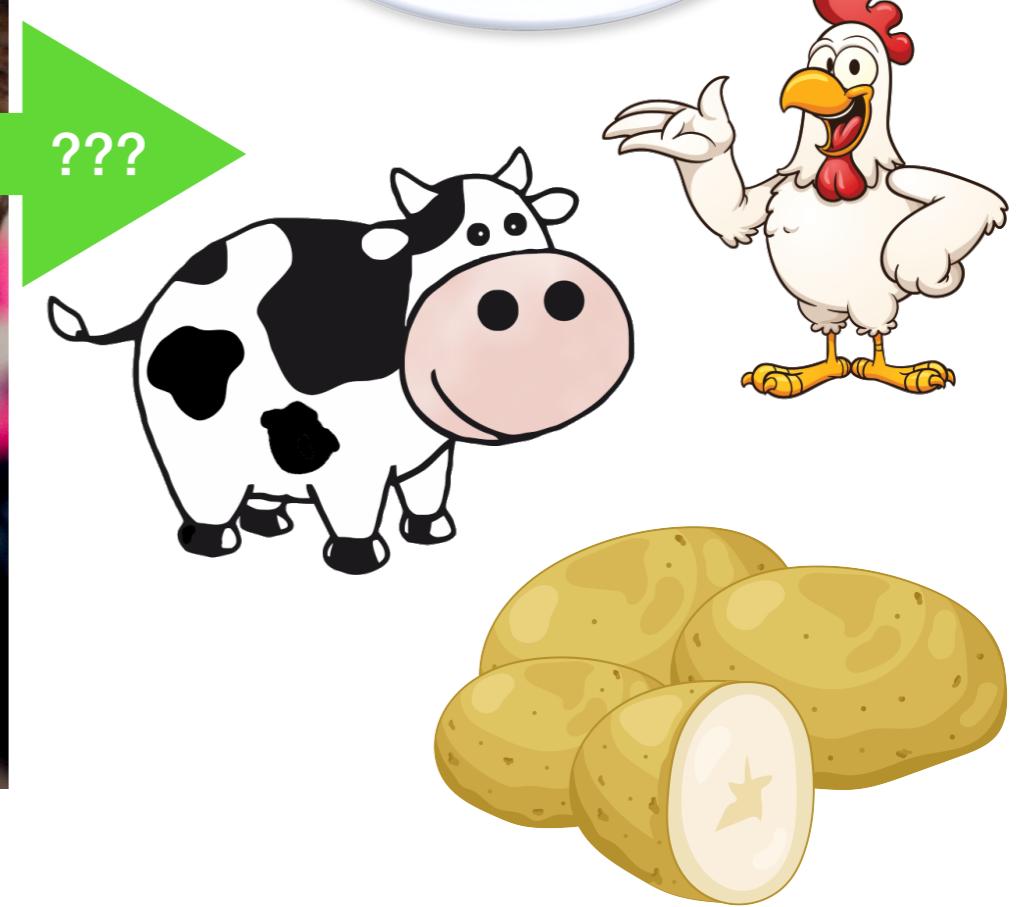
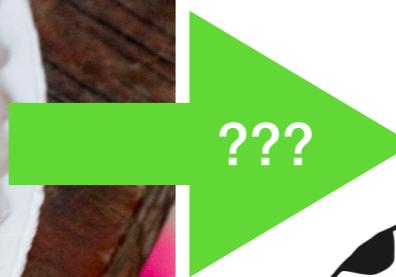
Prof. Seth Lloyd



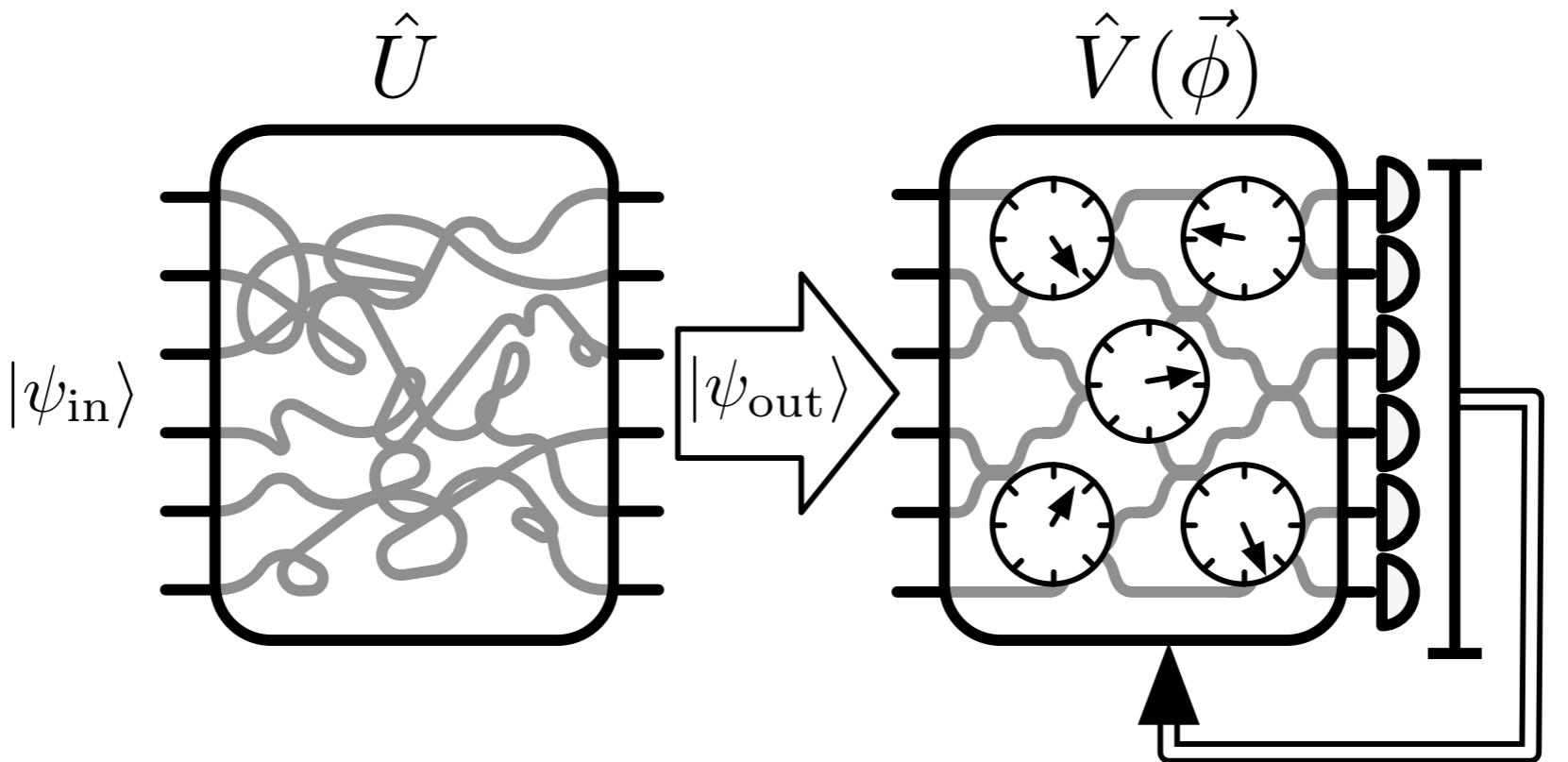
Variational Learning

“Can we unravel the action of an unknown unitary on a known input state, to learn the inverse of the black box operation which efficiently returns Ψ_{in} dynamics?!”





Variational Learning



$$L(\vec{\phi}) = 1 - |\langle \psi_{\text{in}} | \hat{V}(\vec{\phi}) | \psi_{\text{out}} \rangle|^2$$

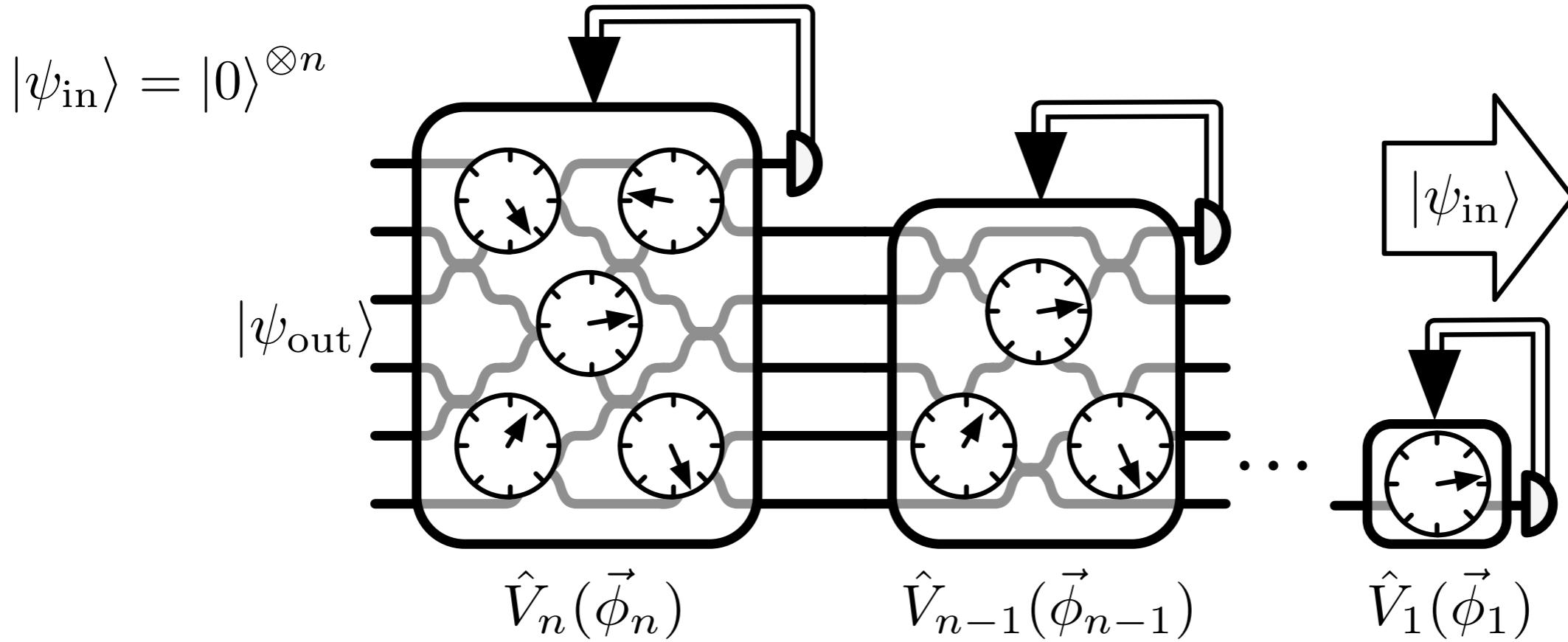
$$\min_{\vec{\phi}} L(\vec{\phi}) = 0$$

↑
1/D

$$\rightarrow \hat{V}(\vec{\phi}) = \hat{U}^\dagger$$

$$\max_{\vec{\phi}} |\langle \psi_{\text{in}} | \hat{V}(\vec{\phi}) | \psi_{\text{out}} \rangle|^2$$

Variational Quantum Unsampling



$$L_1(\vec{\phi}_n) = 1 - |\langle 0_1 | \hat{V}_n(\vec{\phi}_n) | \psi_{\text{out}} \rangle|^2$$

$$L_2(\vec{\phi}_{n-1}) = 1 - |\langle 0_2 | \hat{V}_{n-1}(\vec{\phi}_{n-1}) | \psi_{\text{out}} \rangle|^2$$

⋮

⋮

$$L_n(\vec{\phi}_1) = 1 - |\langle 0_n | \hat{V}_1(\vec{\phi}_1) | \psi_{\text{out}} \rangle|^2$$

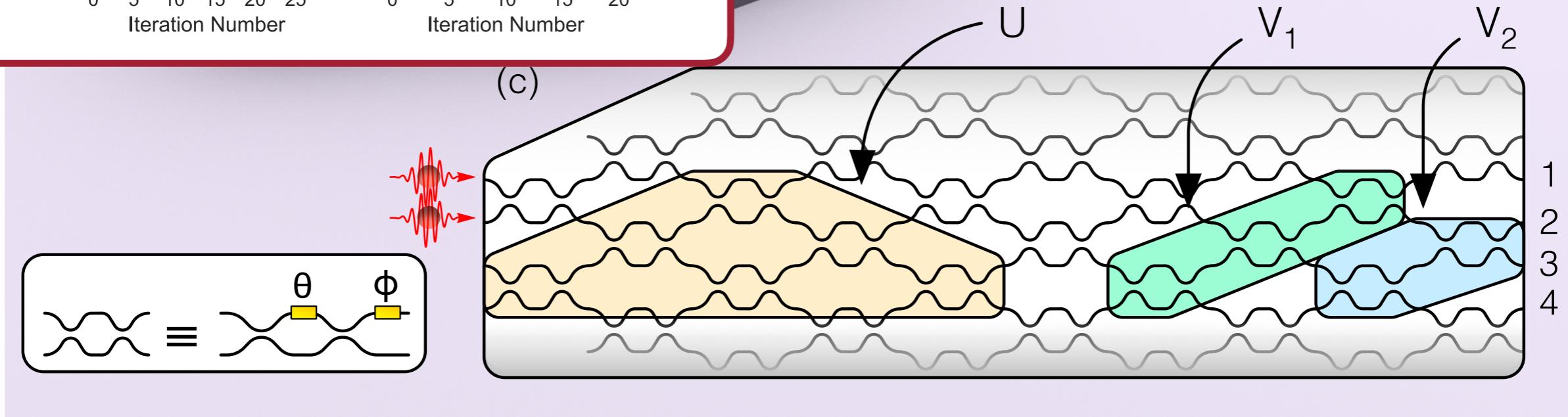
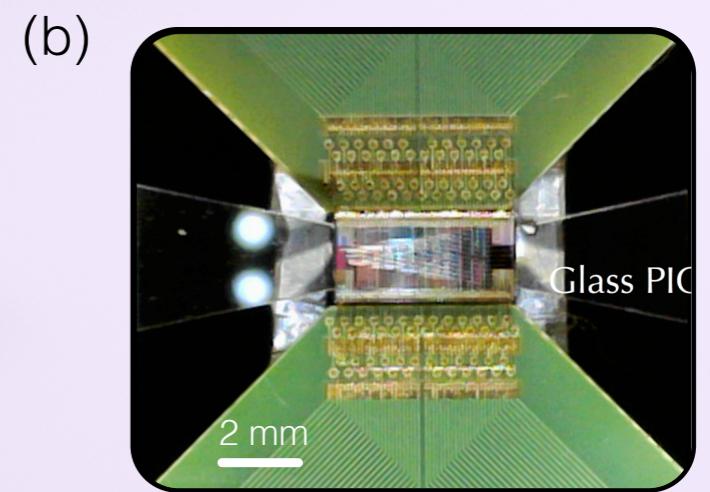
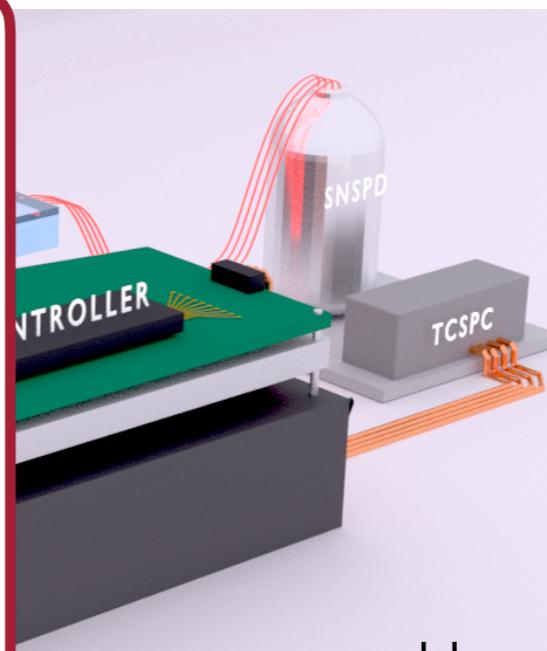
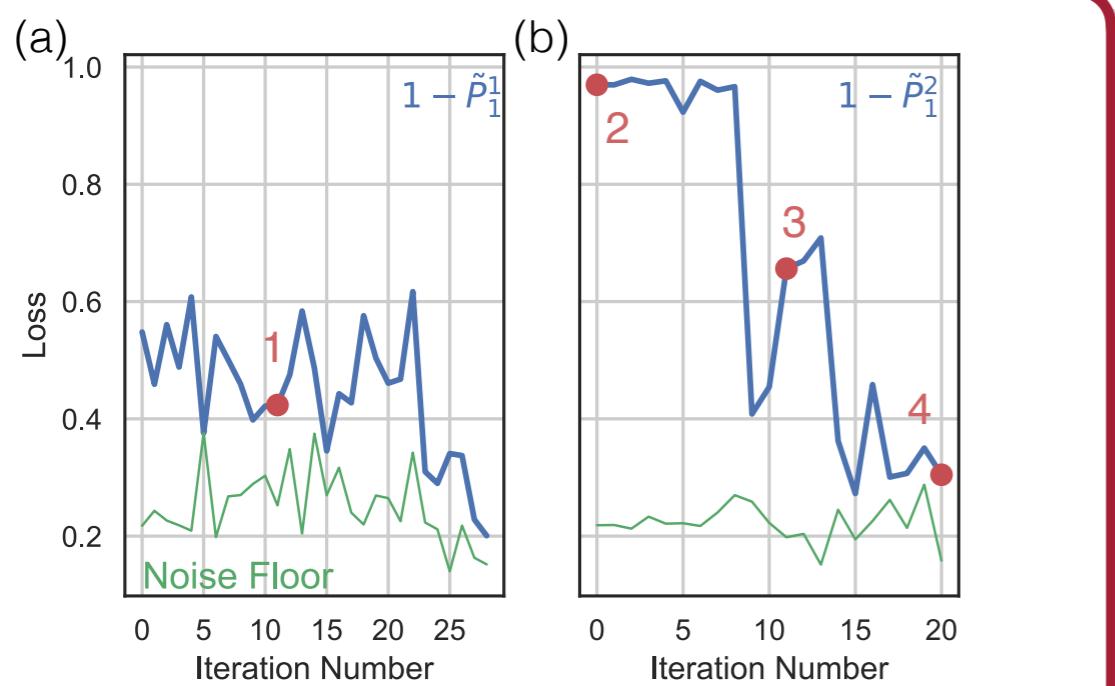
Verification:

$$(1) \text{ Inspect } V \rightarrow \hat{V}_{\text{sol}} = \prod_i^n \hat{V}_i$$

(2) Reduced tomography

$$L_i \approx \mathcal{O}(1)$$

Experimental Boson Unsampling



Photon source

Pairs @ 1582 nm via SPDC
Custom PPKTP Xal¹

Circuit

Si photonic PNP
176 tuneable phase
shifters^{2,3}



Mihika Prabhu



Darius Bunandar



Nick Harris



Franco Wong
Changchen Chen

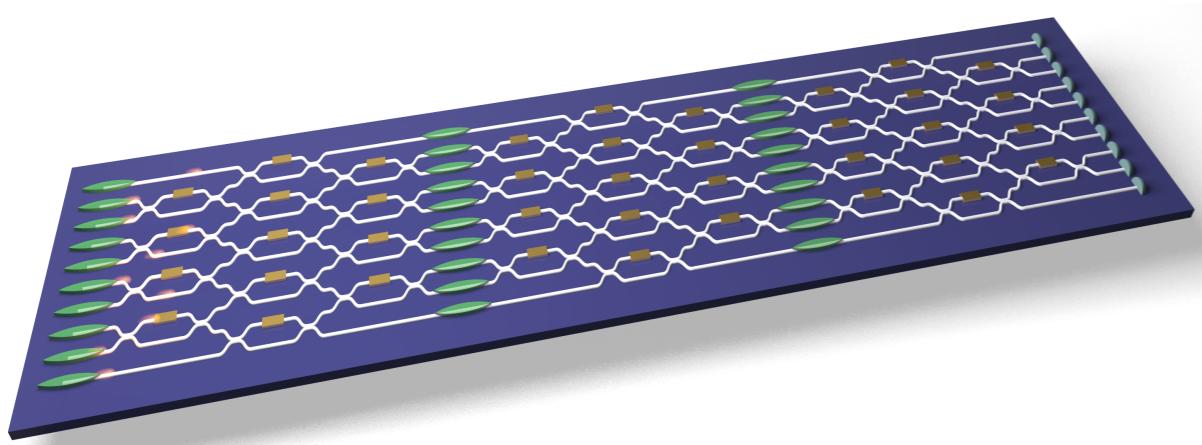
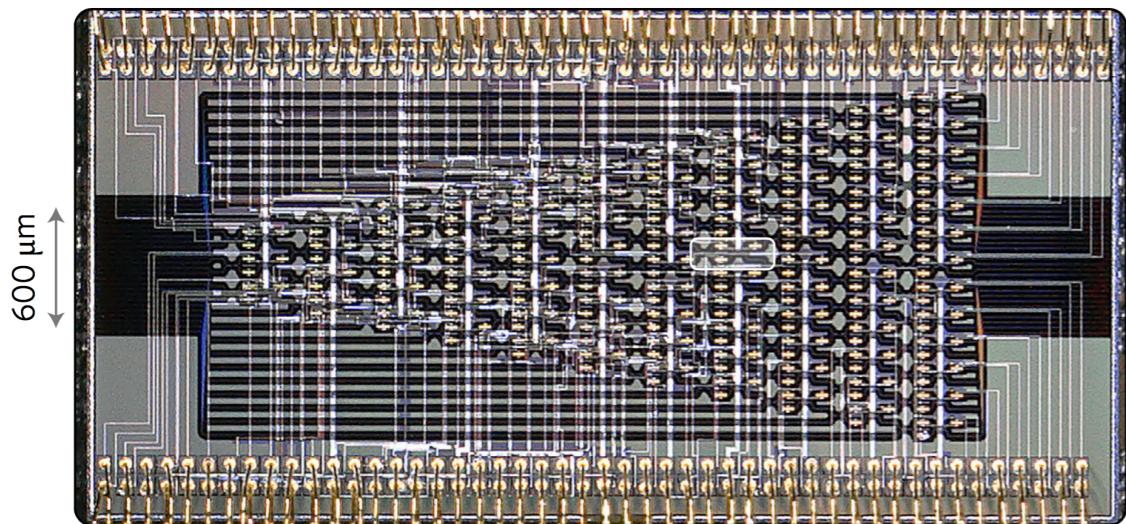
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³N. C. Harris, J. Carolan et al., *Optica* 5, 12 (2018).

The Future...

The Future...



Optical Matrix Processors

Scaling to ~100 of optical modes

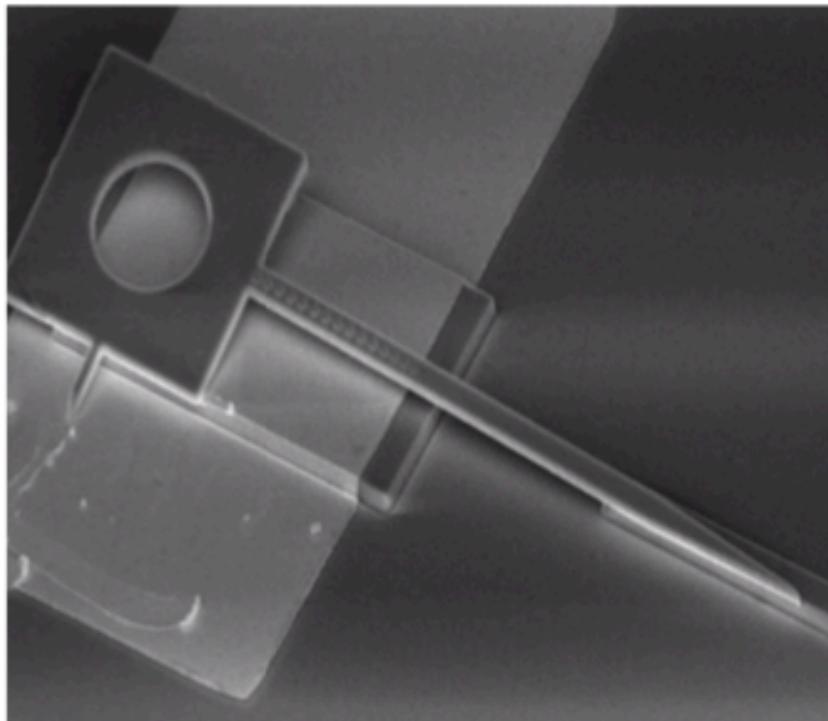
- Optical Machine Learning
- Optical FPGA for signal processing
- New imaging modalities

Quantum Photonic Processors

10's of photons w/ nonlinearities

- Practical molecular simulations
- Quantum machine learning
- New architectures: e.g. recurrent QNN for time sequenced data analysis

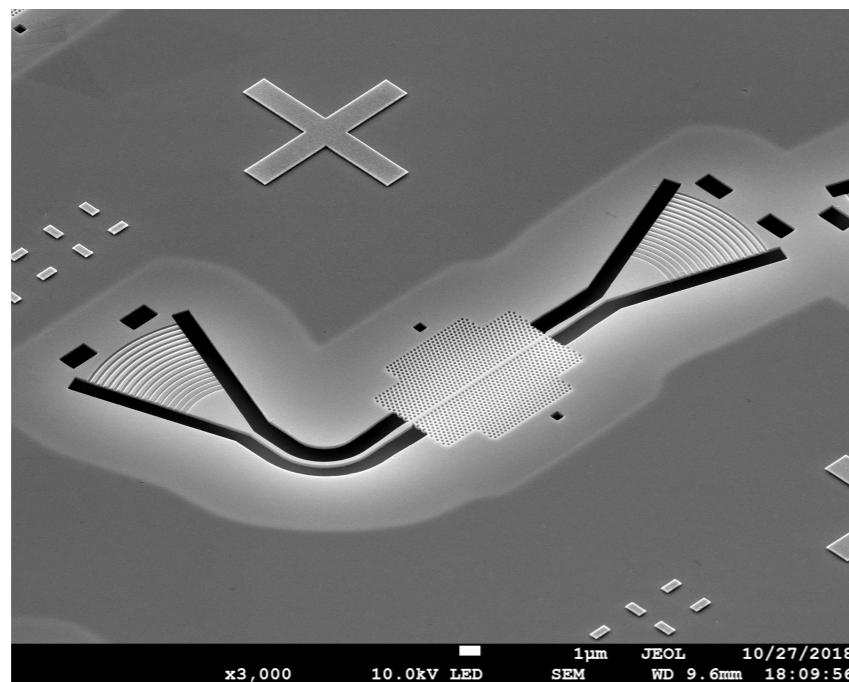
The Future...



Hybrid Integration

New materials

- Phase change materials quasi-static
- EO materials for cryogenic & low loss modulation (LN, BTO)
- High Q/V chiplets for nonlinear optics
- New emitters & quantum memories



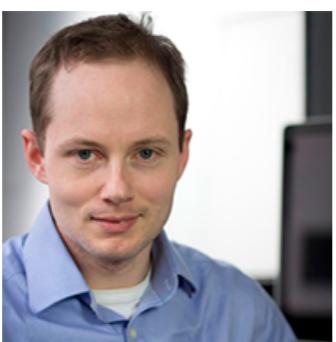
Optical Nonlinearities

Ultra-high cooptivity QD-Cavity

- Quantum repeaters
- Forward error correction
- Universal Quantum Computation

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Academic



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Prof. Seth Lloyd



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NVIDIA®

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Quantum

Photonics

Industry



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Google AI



Dr. Michael Hochberg

ELENION
TECHNOLOGIES



Dr. Jonny Olson

ZAPATA



Dr. Nicholas Harris

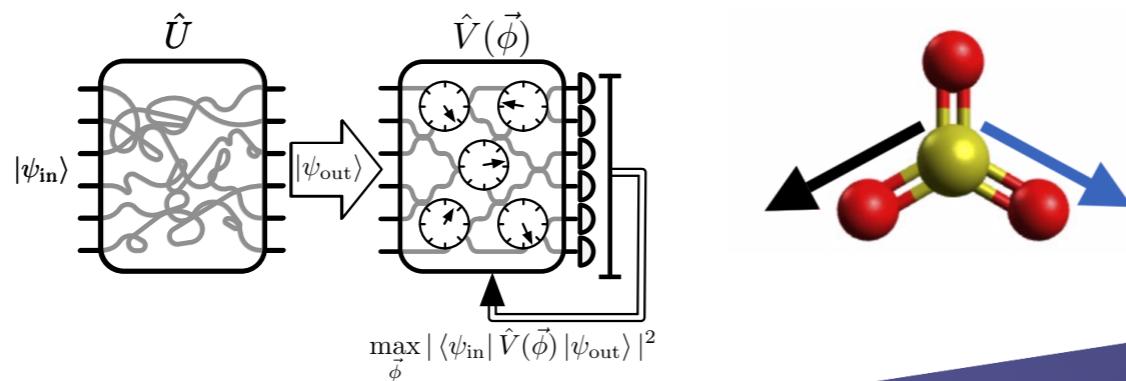
LIGHTMATTER



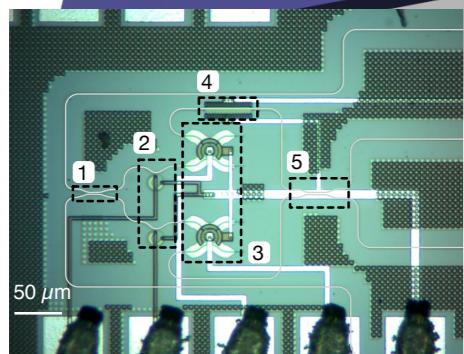
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Thanks!

Applications

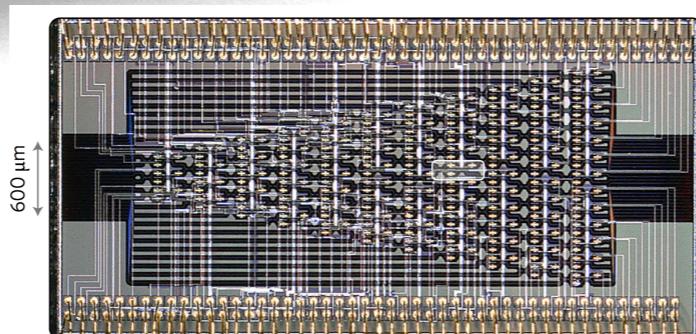


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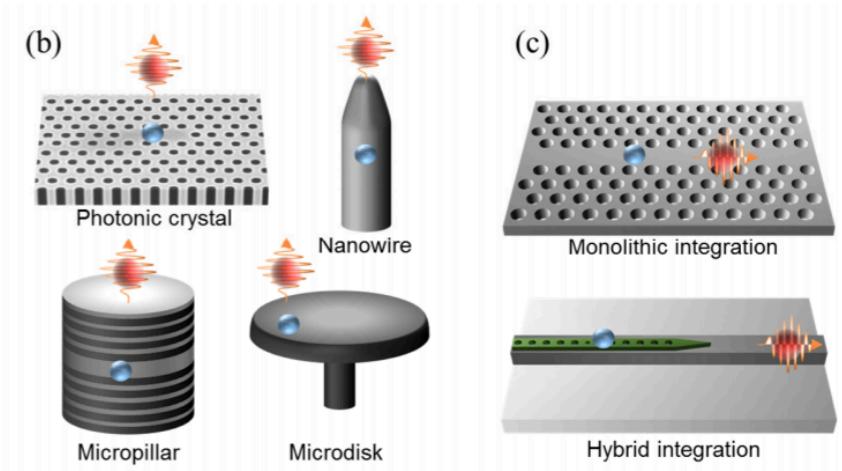
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Nonclassical
states of light



J. Carolan et al., *Science* 349, 711 (2015)
 N. C. Harris, J. Carolan et al., *Optica* 5, 12 (2018)

Reconfigurable
optical circuitry



J. Kim et al., *Optica* 4, 291 (2020)

Nonlinear
light-matter interaction

Thanks!

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