

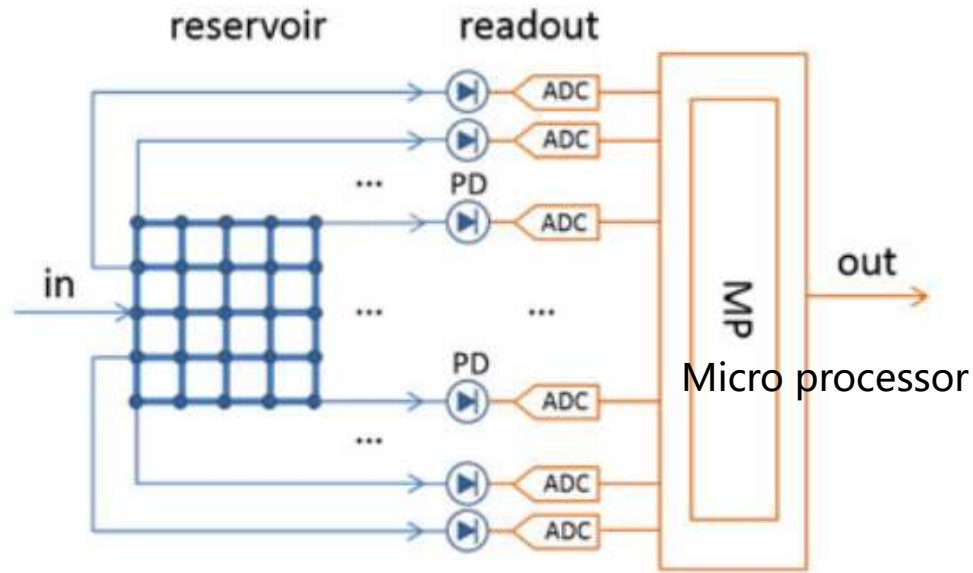
深度学习和物镜调研

畅星兆, 2019.09.01

Photonics AI 训练时间

- 结论：
 - 目前阶段，Photonics AI 仍然使用电路进行模型参数的计算
 - 训练时间与传统计算机训练时间完全相同
- 优势
 - 运行速度快，支持更快的input stream
 - power efficiency

水库计算

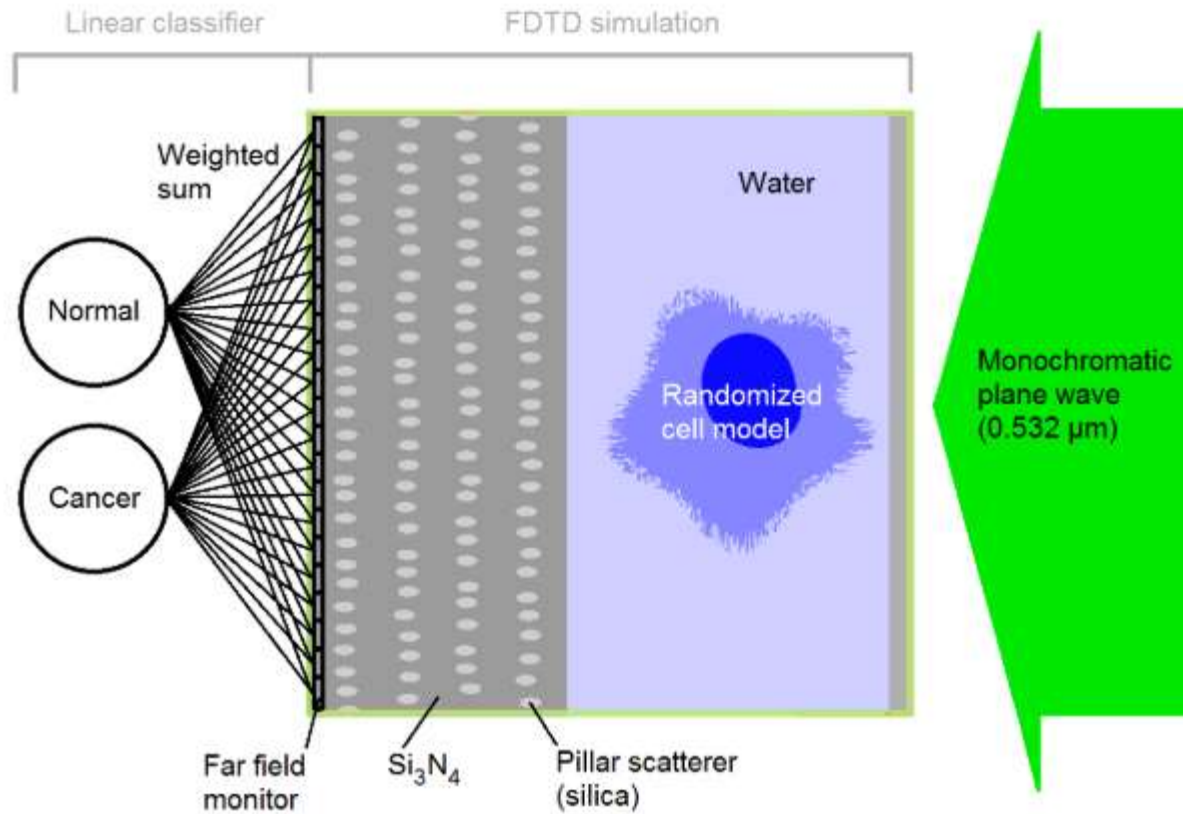


(Katumba et al., Neuromorphic Computing Based on Silicon Photonics and Reservoir Computing, IEEE Journal of Selected Topics in Quantum Electronics, 2018)

- 强调new paradigm, analog computing,
- Readout: machine learning on digital computers.
- We show **how** passive reservoir computing chips **can be used** to perform a variety of tasks (bit level tasks, nonlinear dispersion compensation, etc.) at high speeds and low power consumption.

- 训练方法 (Optical Readouts)
 - separate high-photodetectors
 - photodetectors tend to be costly due to footprint
 - can only measure the intensities
 - trains the weights based on simulation
 - fabrication tolerances
 - pretraining-retraining
 - fabrication tolerances
 - (preferred)single high-photodetectors
 - does take some time
 - still requires external microprocessor
- 使用与传统方法一样的机器学习算法
 - applied in the readout system to teach computer systems how to perform (or interpretate) complex tasks.
 - calculated in the electrical domain

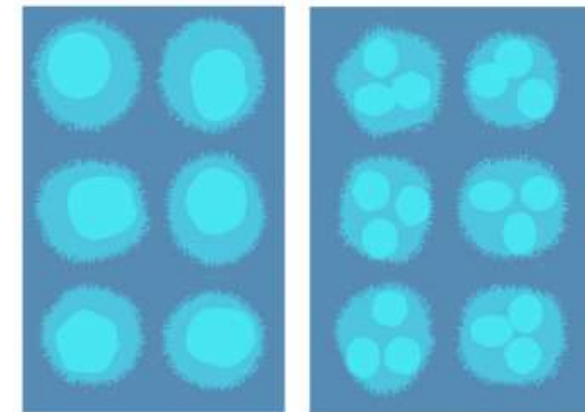
Pillar scatterer paradigm



- monochromatic plane wave
- pillar: randomly generated, kept fixed
- simulation: cells and pillar scatterer (by setting some rules)
- fast and energy efficiency: when running models (not training)
- application: high throughput label-free cell sorting

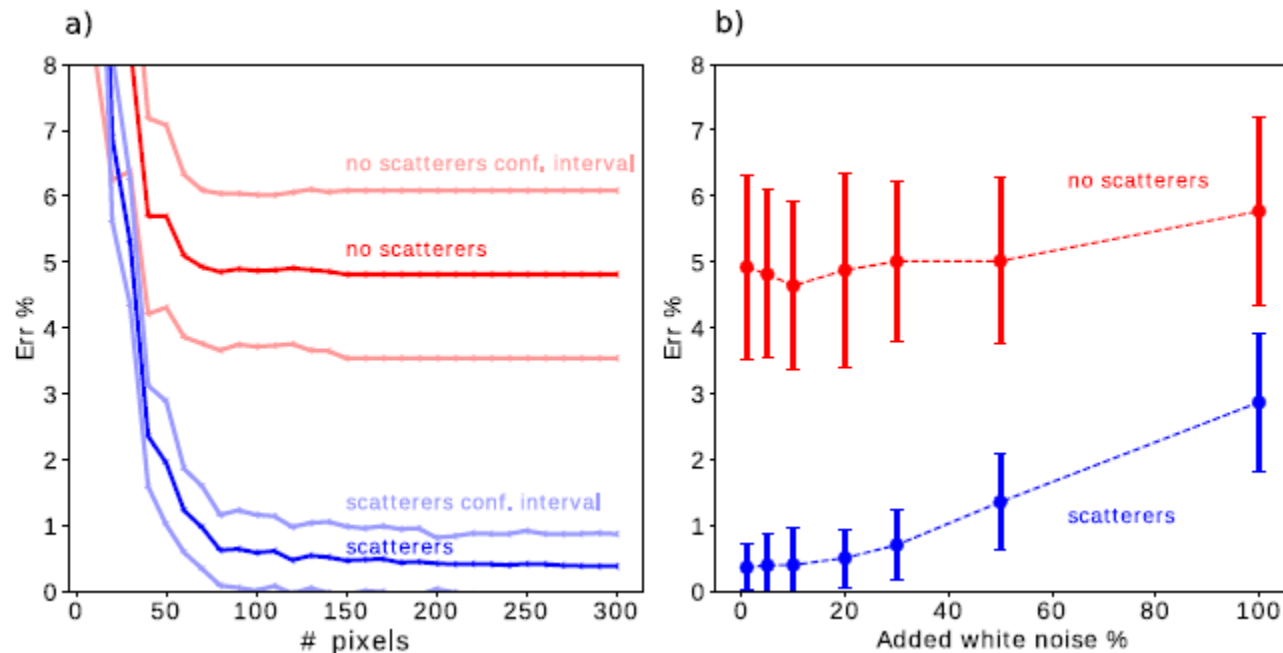
Normal cells and cancer cells form two clusters after calculating weighted sum.

(Alessio Lugnan et al., Integrated pillar scatterers for speeding up classification of cell holograms, Optics Express, 2017)



generated "normal" and "cancer" cell

Result

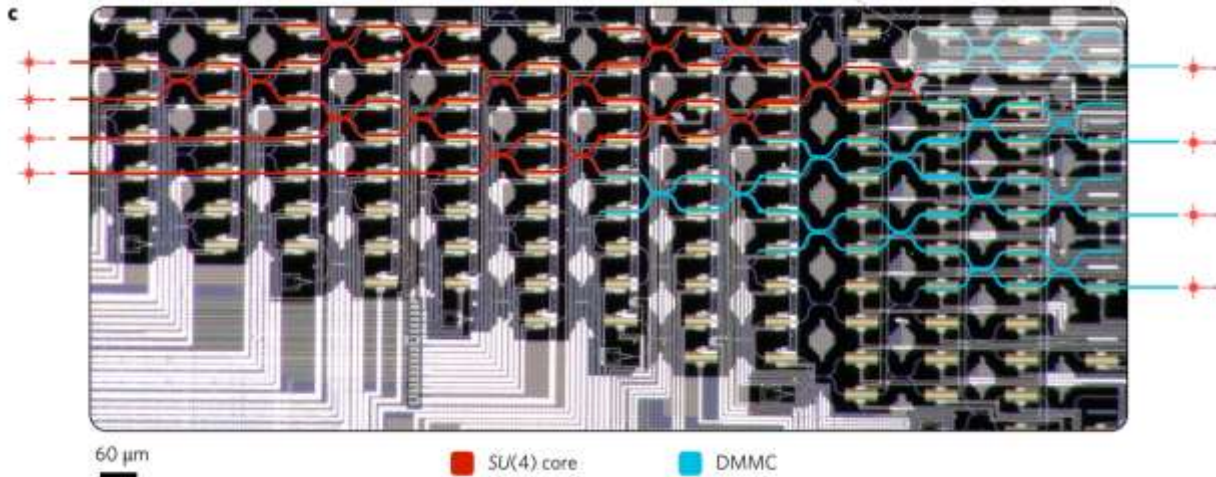


- 参数: 4 layers, $A_r = 150$ nm, $D = 2.85$ μ m, UV (337.1 nm) laser source
- a considerable error rate reduction
- classification performance increase more than 50%

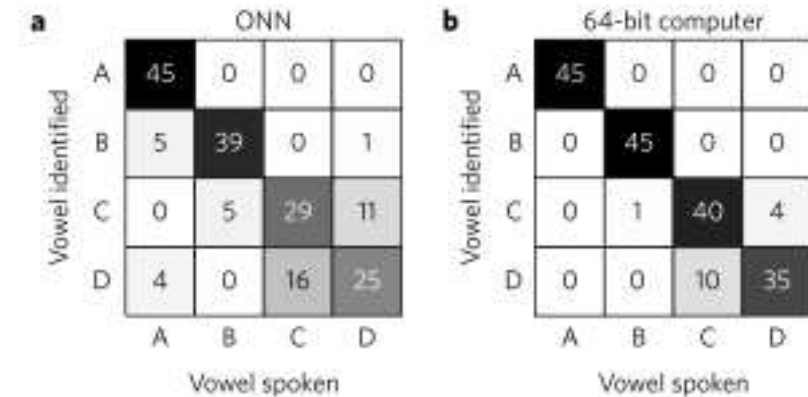
仿真结果证明添加柱状反射结构后，对于该系列图片的识别错误率明显下降。

- 问题: 1. 添加该结构是否对分类任务有帮助?
2. 添加该结构是否只对特定分类任务有帮助?

Deep Learning with coherent nanophotonic circuits



- Featuring a cascaded array of 56 programmable Mach-Zehnder interferometers
- Power efficiency
- Vowel recognition



comparable results

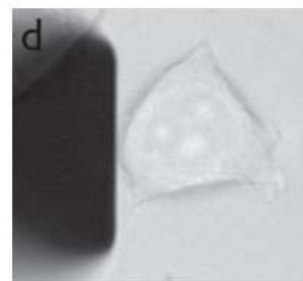
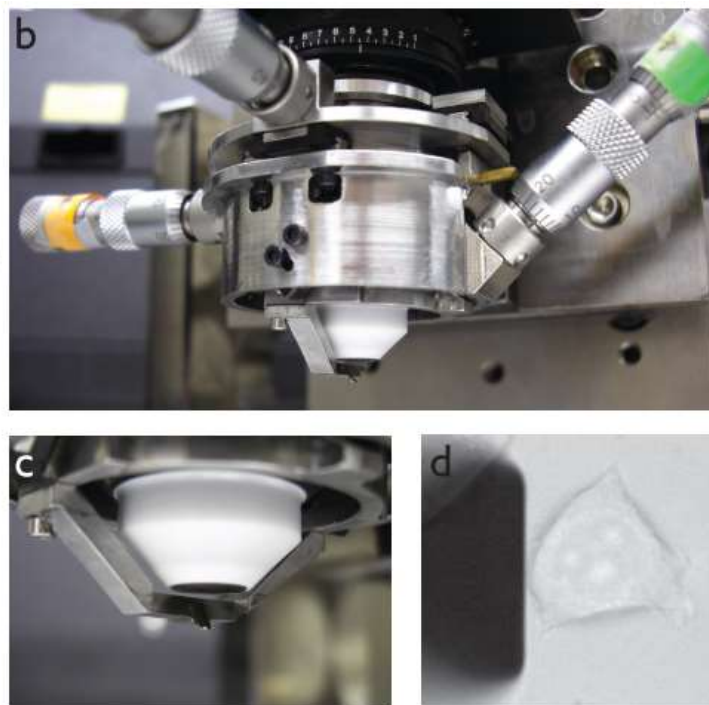
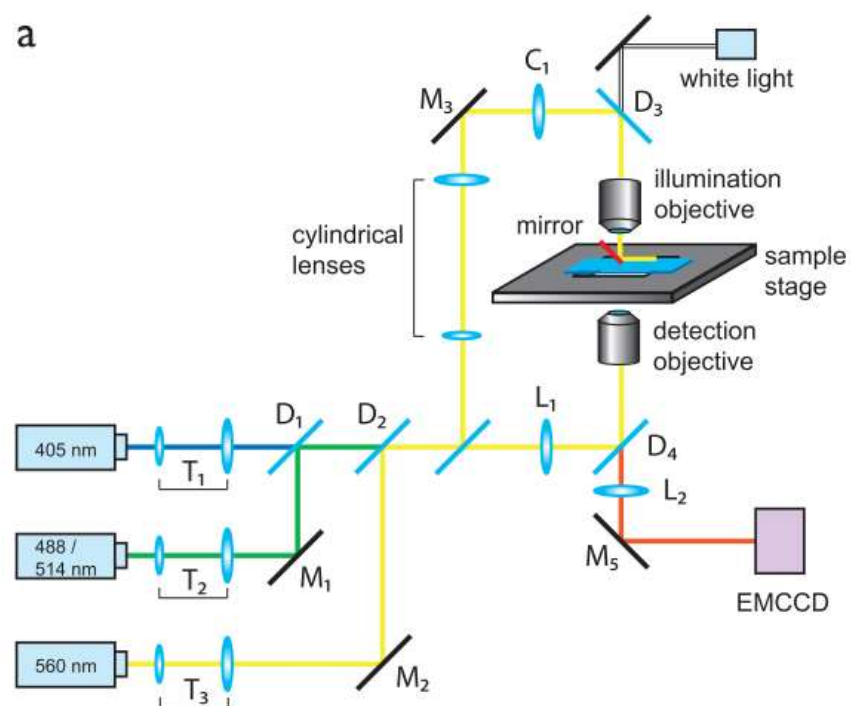
(Shen Yichen et al., Deep learning with coherent nanophotonic circuits, Nature Photonics, 2017)

We trained the matrix parameters with the standard back-propagation algorithm using a stochastic gradient descent method **on a conventional computer.**

调研：翻折物镜

- Thorlab调研
 - 网站：未找到
 - 销售代表：未找到
- Olympus调研
 - 网站：未找到
 - 销售代表：正在整理（显微镜物镜切换盘？）

反射镜



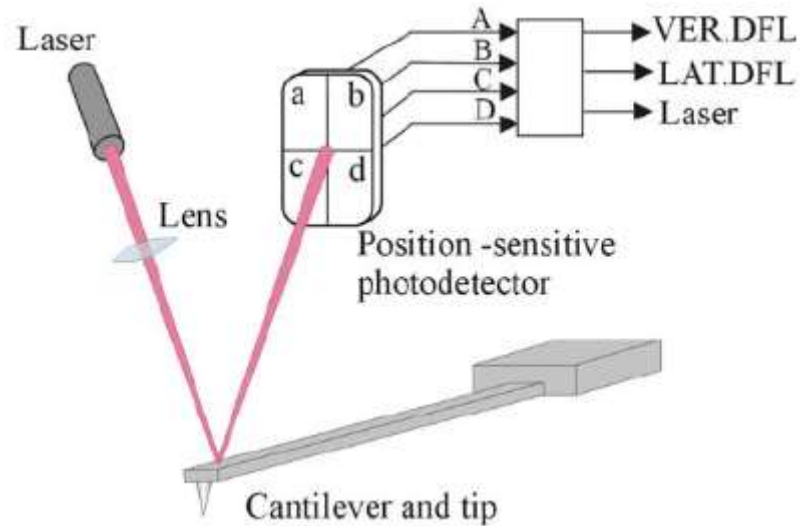
a: 示意图, b: 实拍图, c: 近照, d: 宽场图像

(Gebhardt et al., Nature Methods, 2013)

- 组成：自制的xyz位移台+AFM悬臂 (AFM cantilever)
- 自制的xyz位移台：4个不锈钢圆筒，第一个固定在物镜上，其余三个级联，分别用于控制悬臂在三个维度上运动。
- AFM悬臂：通过镀膜形成反射镜。

AFM Cantilever

- AFM原理图



(<https://wenku.baidu.com/view/ad63ad0b01f69e31433294db.html>)

- 文中: HYDRA2R-100N-TL-10, Nanoscience (not found)
- 处理: (custom-coated, thermal evaporation) 1 nm Ti layer followed by 40 nm Al layer