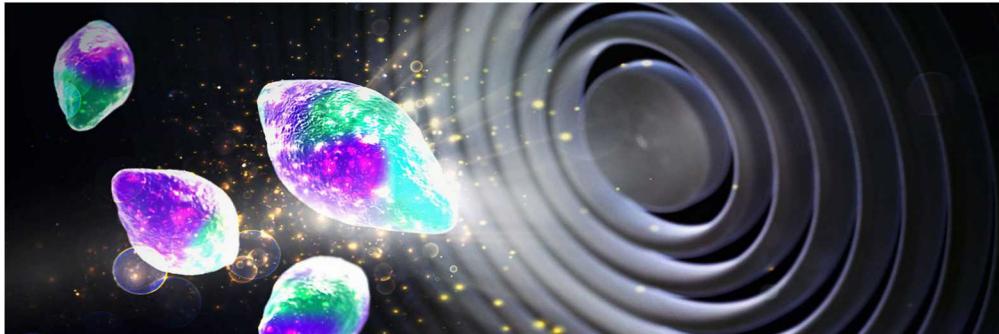


# Electrochemical ion insertion for energy storage and neuromorphic computing



PRESENTED BY

Yiyang Li

This work is supported by the Sandia National Laboratories Truman Fellowship Program, which is funded by the Laboratory Directed Research and Development (LDRD) Program. Sandia National Laboratories is a multi-mission laboratory managed and operated by National Technology and Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA-0003525.



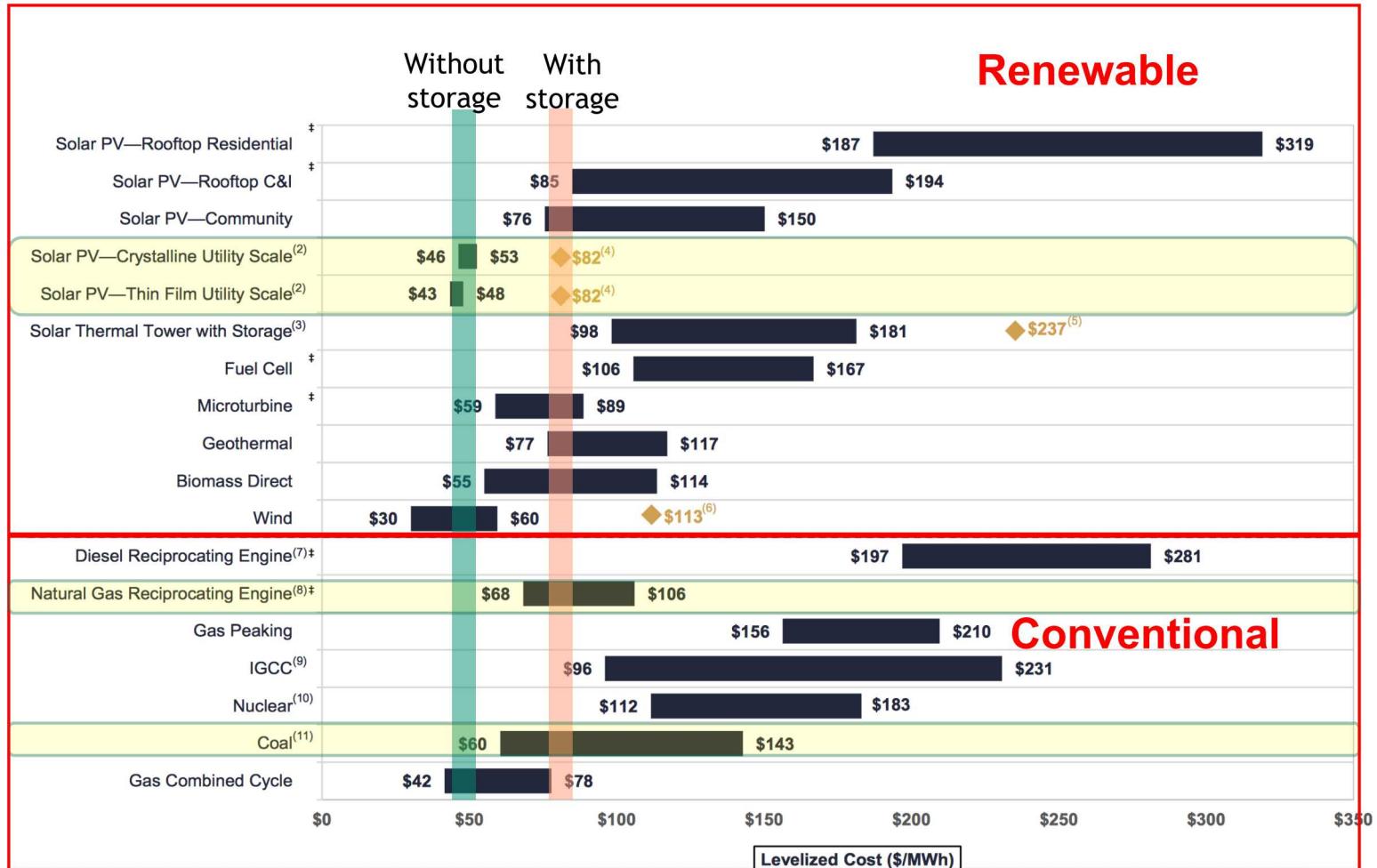
Sandia National Laboratories is a multimission laboratory managed and operated by National Technology and Engineering Solutions of Sandia LLC, a wholly owned subsidiary of Honeywell International Inc. for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.



7 billion people  
16 Terawatt of Power Consumed



# Renewable electricity is cost-effective today



Storage: the key to enabling renewable energy to be used when and where we use it

Source: Lazard



# Energy storage in personal transportation

Honda Clarity, Electric



Honda Accord, Gasoline



90g CO<sub>2</sub> /mile  
\$0.03 / mile

300g CO<sub>2</sub> / mile  
\$0.10 / mile

Electricity: \$0.13/kWh, 300g CO<sub>2</sub> /kWh (SF Bay area), 4 miles/kWh  
Gasoline: \$3.00/gallon, 9000g CO<sub>2</sub>/gallon, 30 miles/gallon

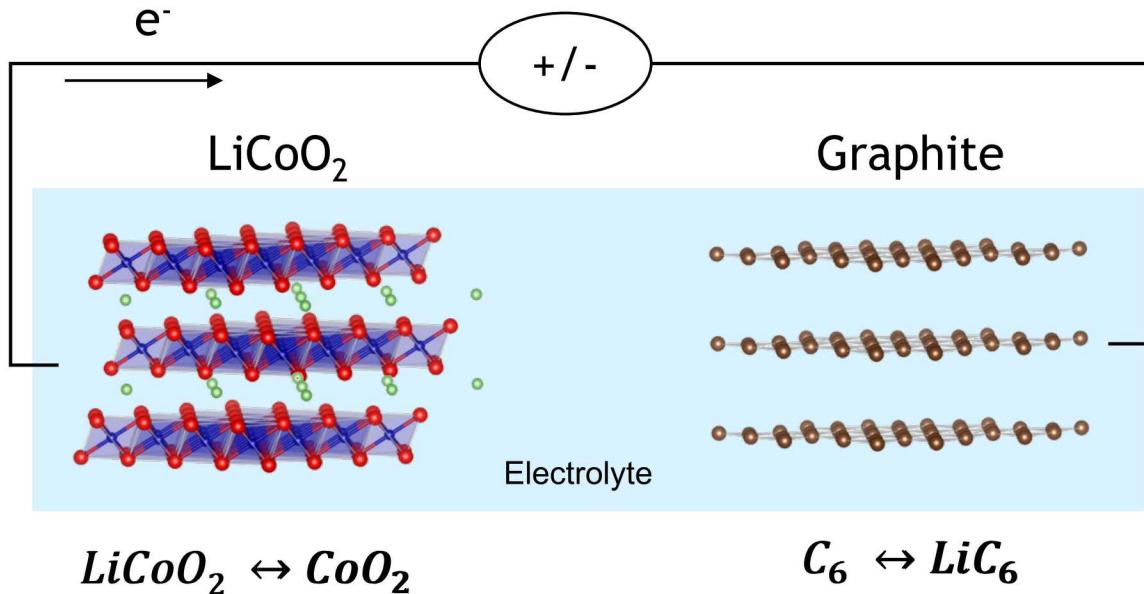
Source: Honda, Tesla





# Electrochemical ion insertion

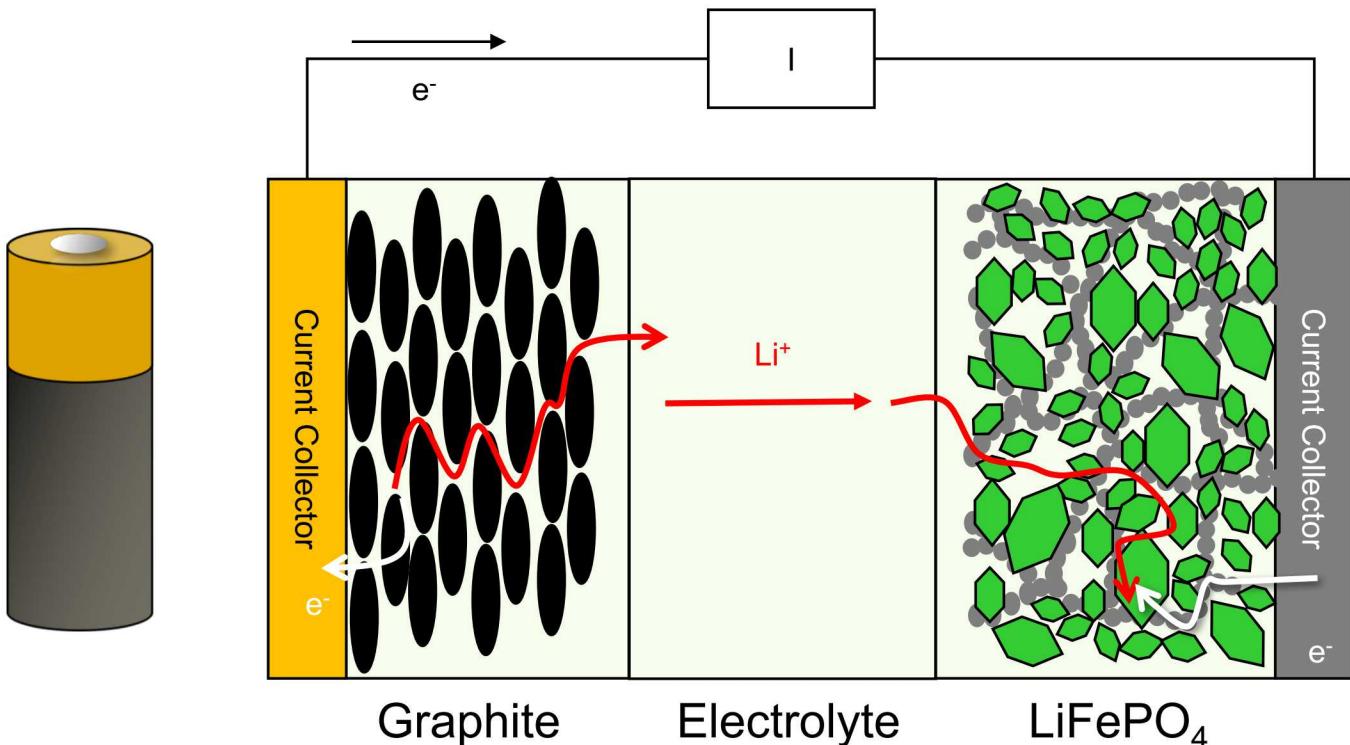
Using electrical signals to dynamically control material composition



Modulating the material composition through current and voltage.  
Alter solid-state chemistry through electronic signals



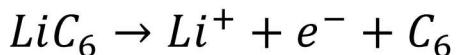
# Ion insertion in a Li-ion battery



$$Energy = QV$$

$$Q = e n_{Li}$$

$$V = \frac{\Delta G_{rxn}}{e}$$



$$\Delta G = -320 \frac{kJ}{mol}$$

$$V = -\frac{\Delta G}{nFe} = 3.4 V$$

Discharge: inserting Li into  $Li_xFePO_4$

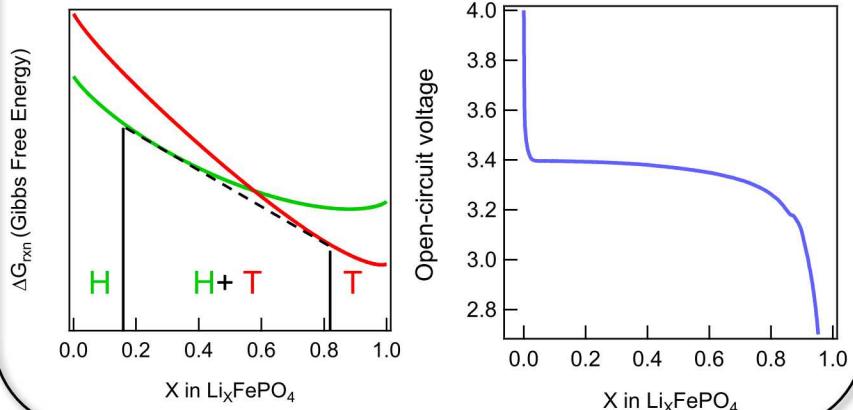
Charge: removing Li from  $Li_xFePO_4$

C-rate: current to (dis)charge in 1 hr

2C: 30 min (dis)charge

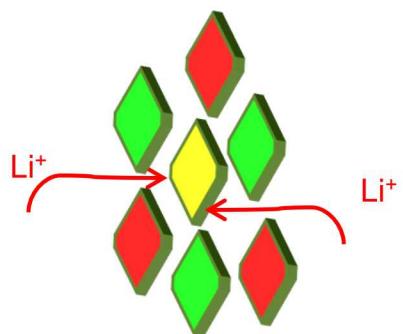
# Phase separation in $\text{Li}_x\text{FePO}_4$

## Large miscibility gap

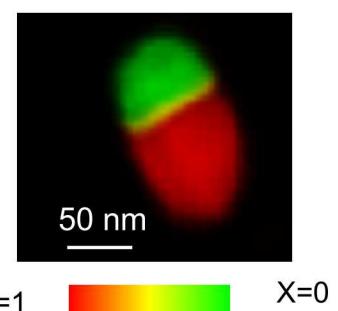


## Phase separation

Inter-particle



Intra-particle

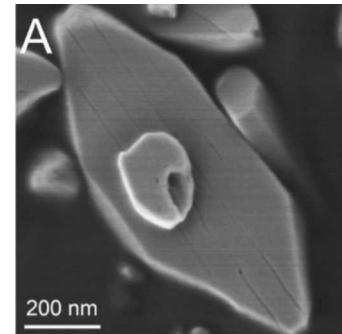


Y. Li et al. Nature Mater. 13, 1149-56 (2014)

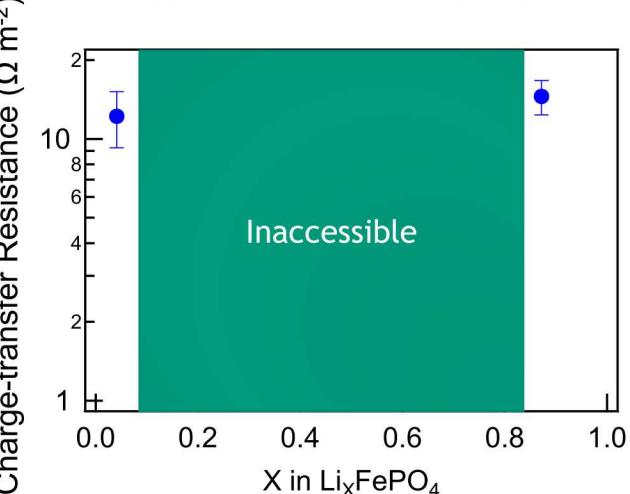
## Consequences of phase separation

### Mechanical strain

Yu et al. *Nano Lett.* (2015)



### Inaccessible stoichiometries

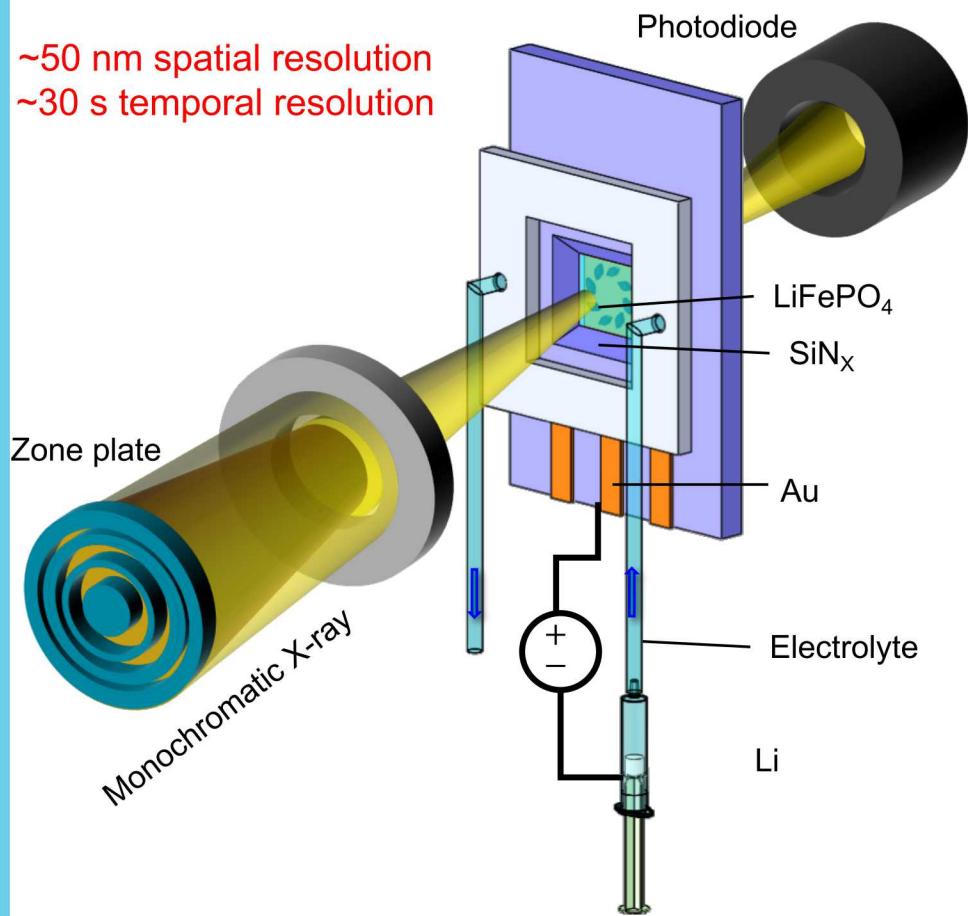


Goal: identify how to control phase separation

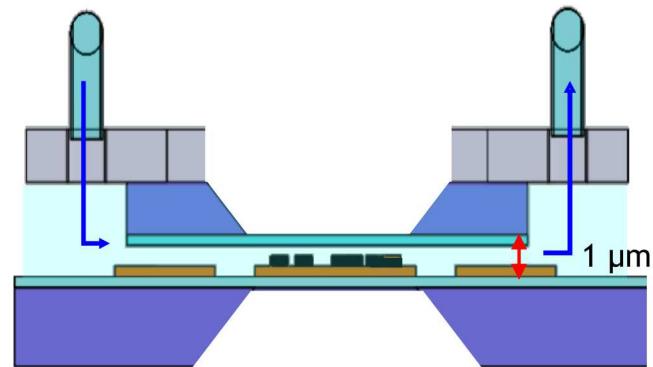


# Visualizing battery (dis)charge using in situ X-ray microscopy

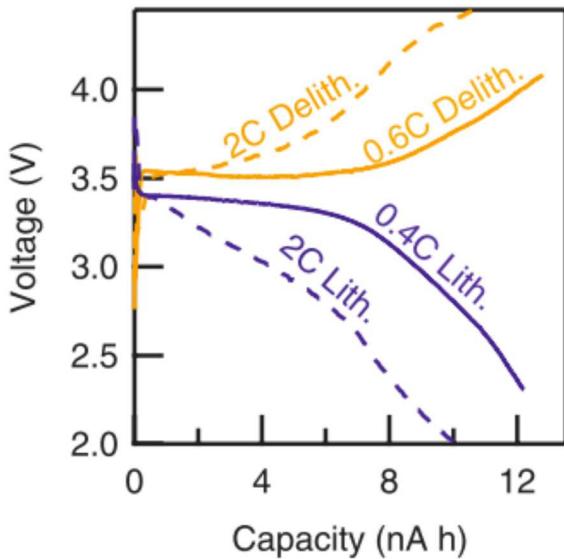
## Tracking lithium insertion during cycling



J. Lim\*, Y. Li\*, et al. *Science*, **353**, 566-571 (2016)  
\*equal contribution authors

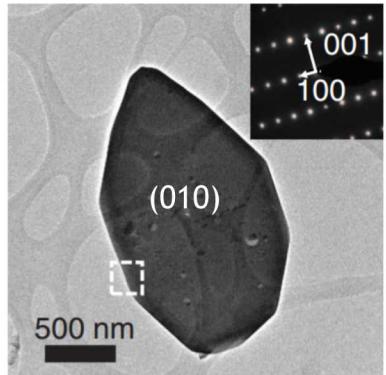


## Robust electrochemical cycling

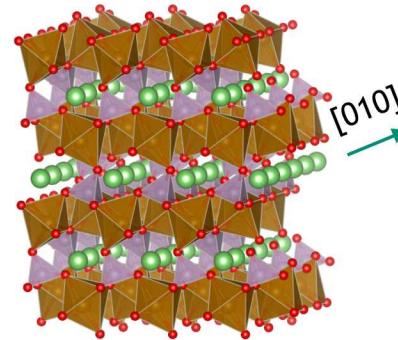




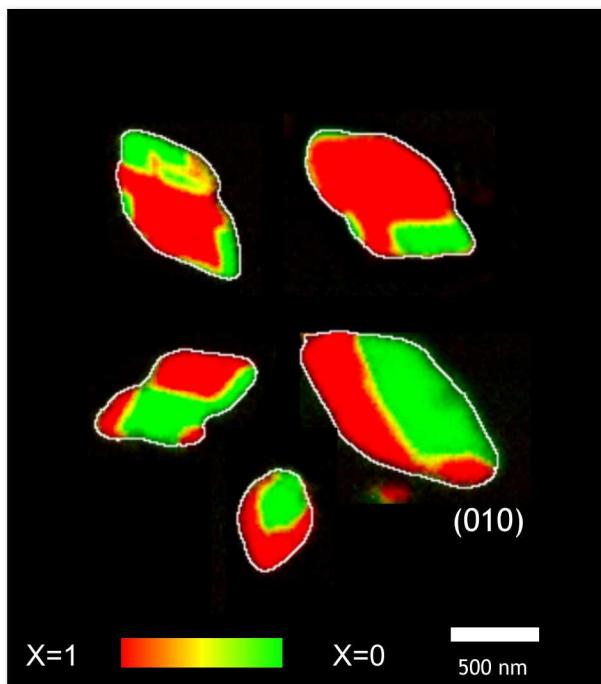
# Tracking lithium insertion within particles *in situ*



100-nm thick single crystalline particles



One dimensional lithium conductor

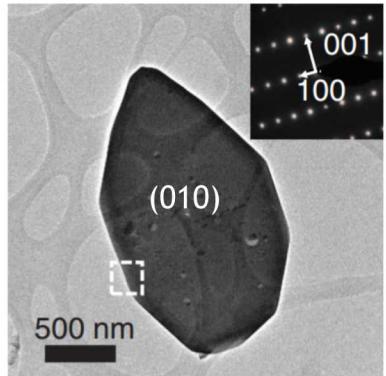


Equilibrium & slow rates

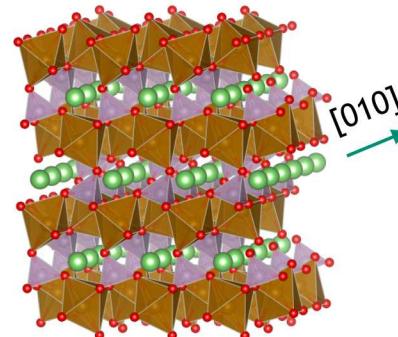
J. Lim\*, Y. Li\*, et al. *Science*, **353**, 566-571 (2016)  
\*equal contribution authors



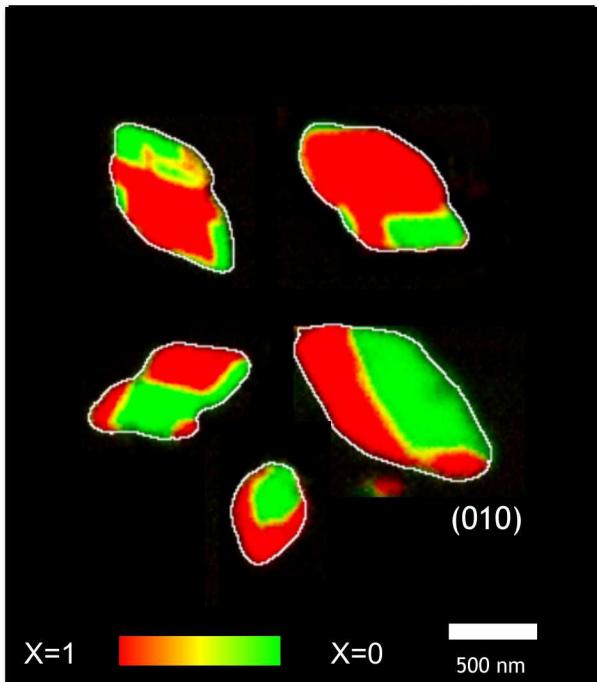
# Tracking lithium insertion within particles *in situ*



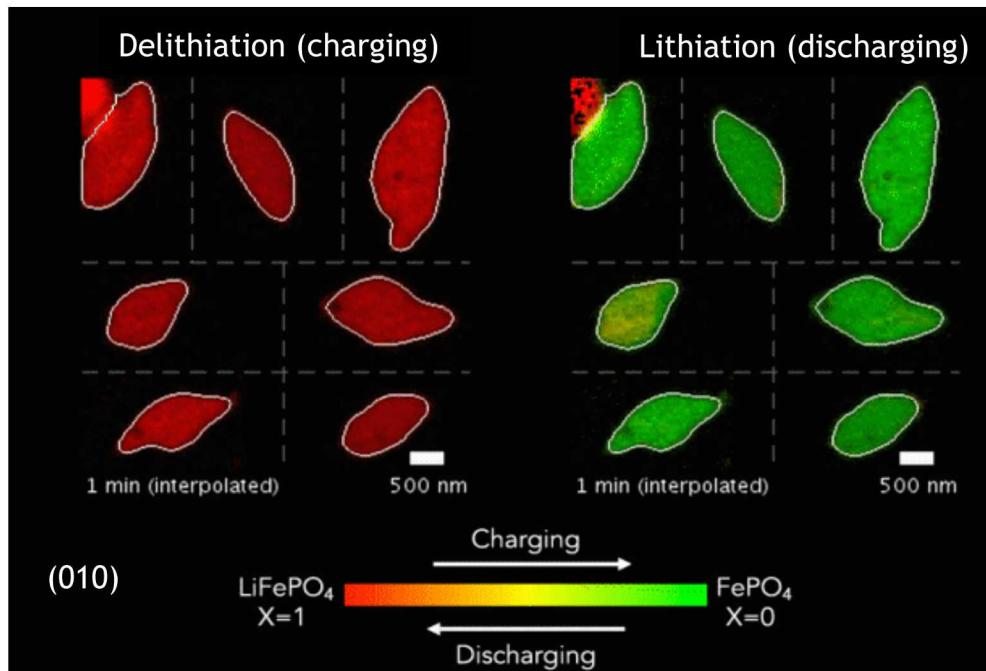
100-nm thick single crystalline particles



One dimensional lithium conductor



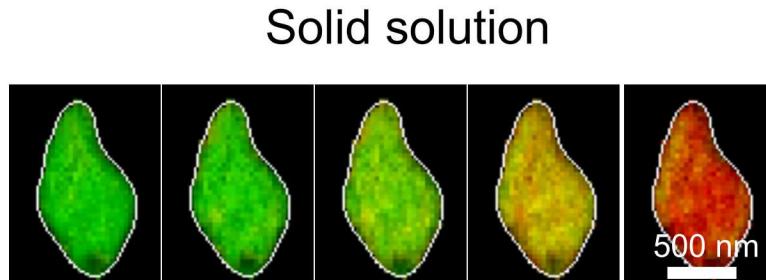
Equilibrium & slow rates



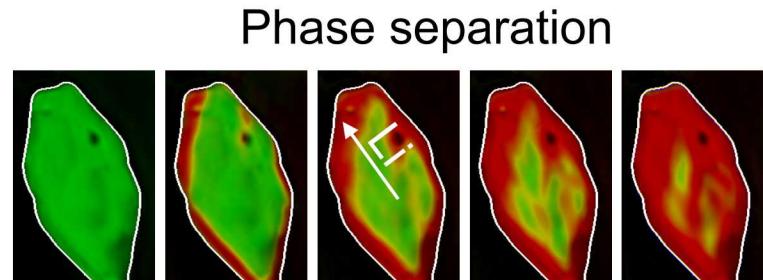
Nonequilibrium lithium insertion



# How do we prevent phase separation?

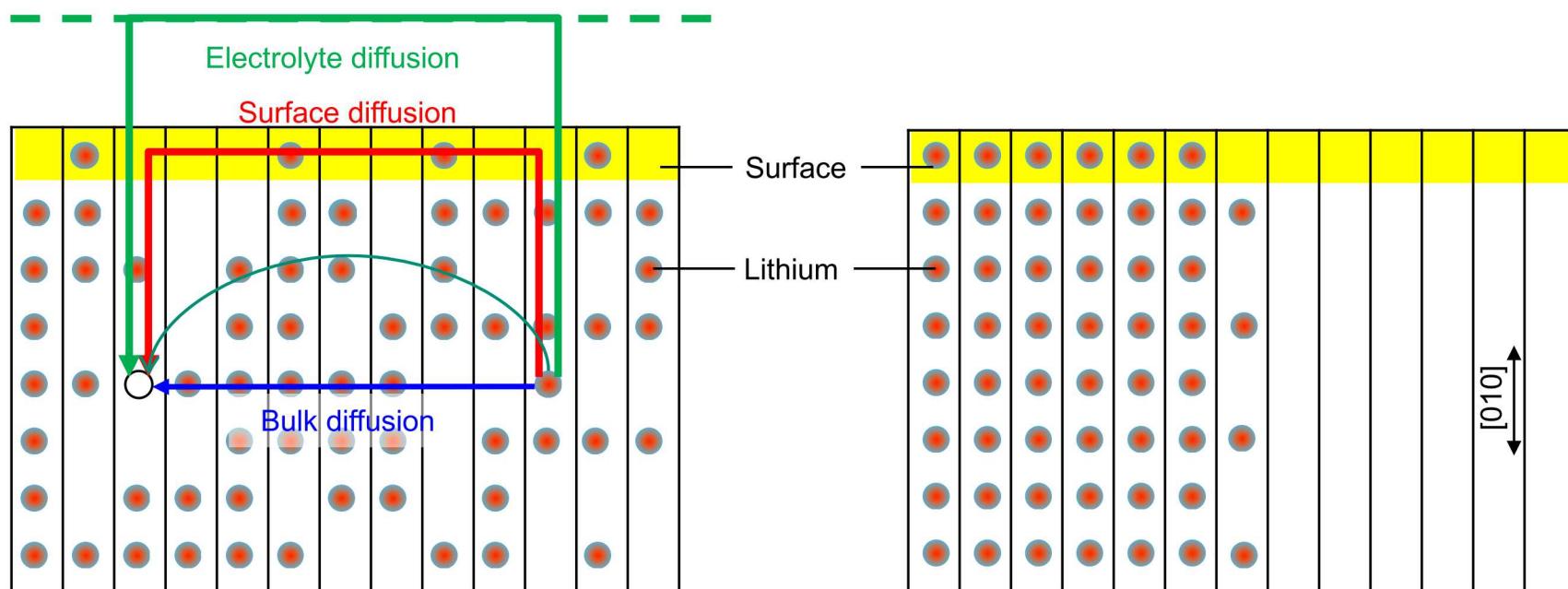


30 min lithiation



7 hour lithiation

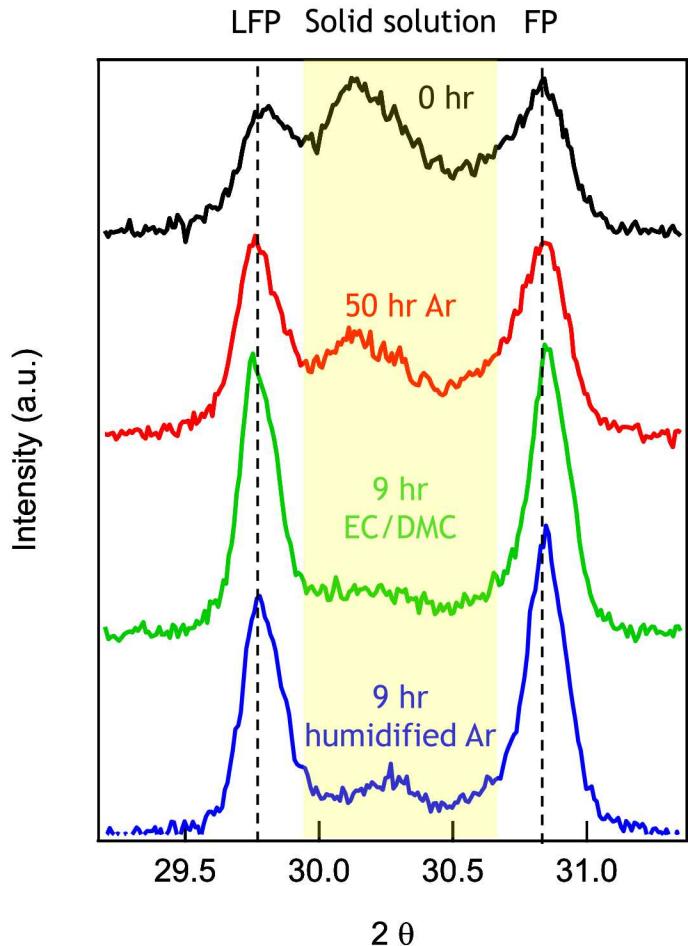
Lithium must diffuse along the non-conducting directions for phase separation



How do we prevent lithium from migrating between channels?

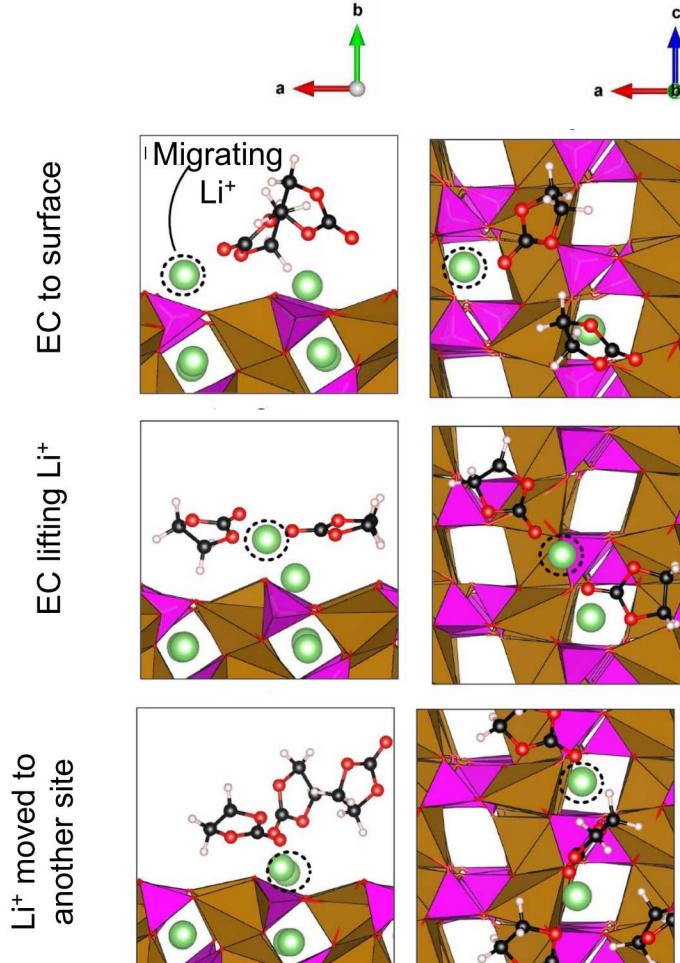


# Lithium migration pathway



Fluid molecules enhance phase separation rate.  
Lithium migrates at the surface

Li, Chen, Islam, Bazant, Chueh et al. submitted



MD shows solvent-assisted lithium migration

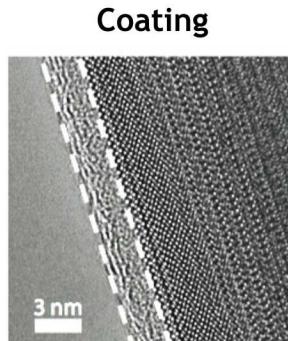
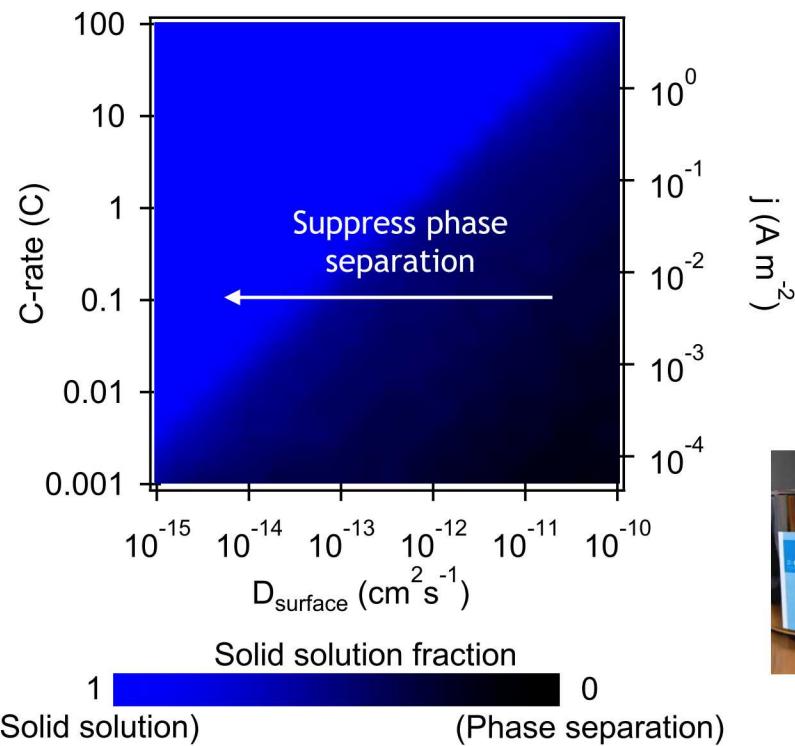
Collaboration: Hungru Chen & Saiful Islam, Univ of Bath



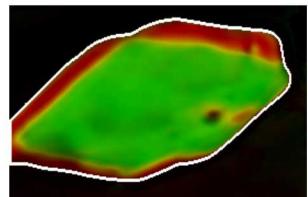
# Surface diffusion controls phase separation

Cahn-Hilliard:  $\frac{\partial X}{\partial t} = \nabla \left[ \frac{D}{k_B T} X(1-X) \nabla \mu_{Li} \right] + j(X, \mu_{Li})$

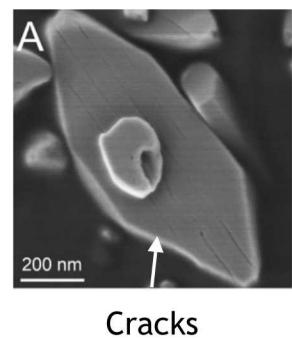
$$\mu_{Li} = \Omega(X)(1-X) + k_B T \ln \frac{X}{1-X} - \kappa \nabla^2 X + \dots$$



Phase boundary:  
2-4% lattice mismatch  
>1 GPa stress

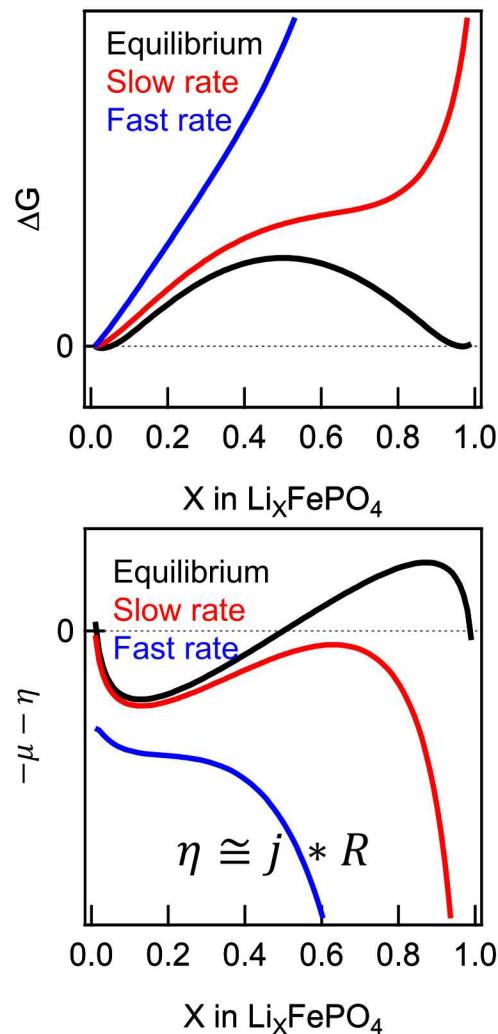
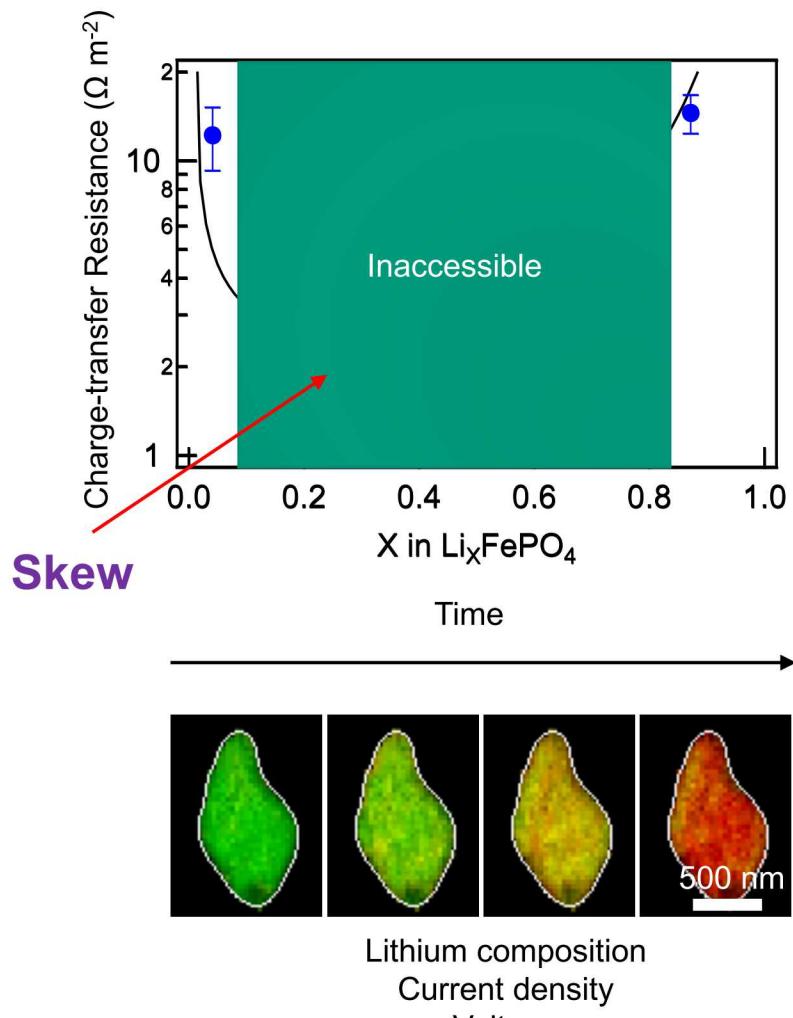


Yu et al. *Nano Lett.* (2015)





# Interfacial reaction controls bulk thermodynamics

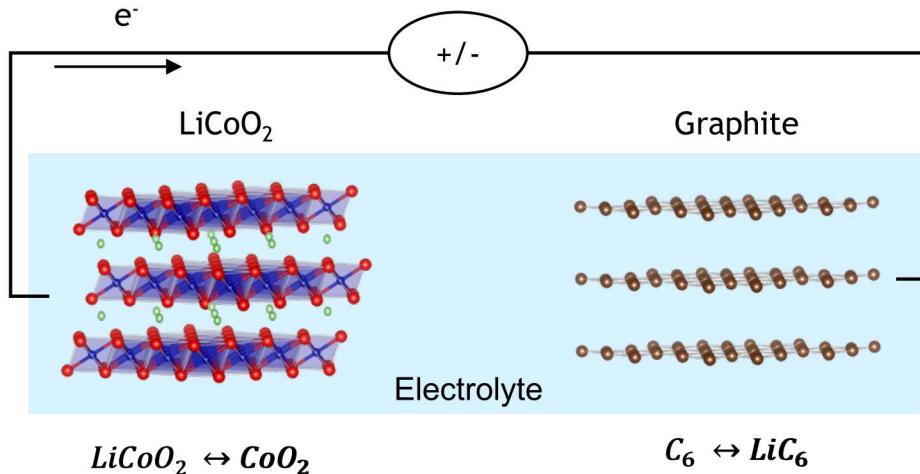


This shape of the resistance suppresses the driving force for phase separation

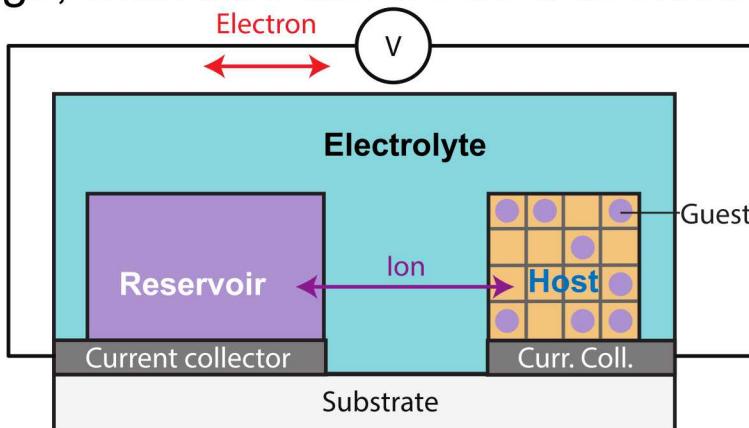


# Electrochemical ion insertion

Modify chemical stoichiometry and conduct solid-state chemistry using current and voltage



Beyond energy storage, what else can we do with electrochemical ion insertion?





# Artificial neural networks and deep learning

Computers are really fast and efficient at task-specific programming

$$\frac{dx}{dt} = x^2$$

```
x=1;  
dt = 0.01;  
for i = 1:1000 {  
    dxdt = x.^2;  
    xnew = dt*dxdt;  
    x = xnew; }  
end
```

Computers struggle when there are no clear instructions for the task

Which one of these images is a cat?



Image recognition, online advertisement, autonomous driving

Artificial neural networks: how can a machine learn to do an operation through repeated training?

Solution: conduct matrix multiplications, while tuning the weights of the matrix

$$\begin{bmatrix} y_1 \\ \vdots \\ y_m \end{bmatrix} = \begin{bmatrix} w_{1,1} & \cdots & w_{1,n} \\ \vdots & \ddots & \vdots \\ w_{m,1} & \cdots & w_{m,n} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

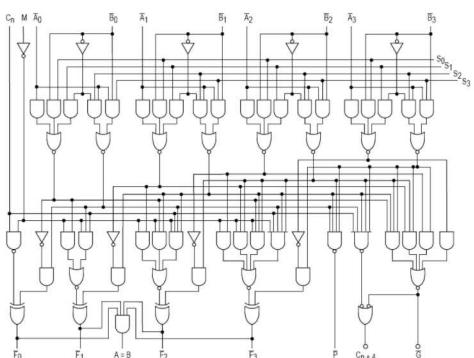


# Digital vs analog computation

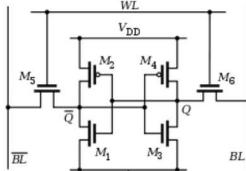
$$\begin{bmatrix} y_1 \\ \vdots \\ y_m \end{bmatrix} = \begin{bmatrix} w_{1,1} & \cdots & w_{1,n} \\ \vdots & \ddots & \vdots \\ w_{m,1} & \cdots & w_{m,n} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

## Digital logic

Arithmetic logic unit for multiplication (read)



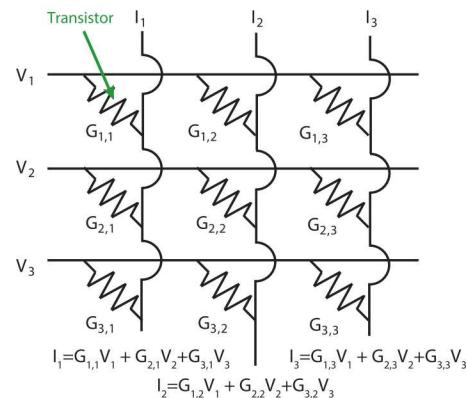
SRAM to store the weights (write)



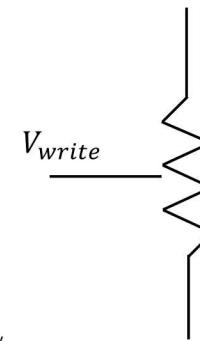
Uses established CMOS technology

## Analog logic

Crossbar for matrix multiplication



Modulate conductance during training



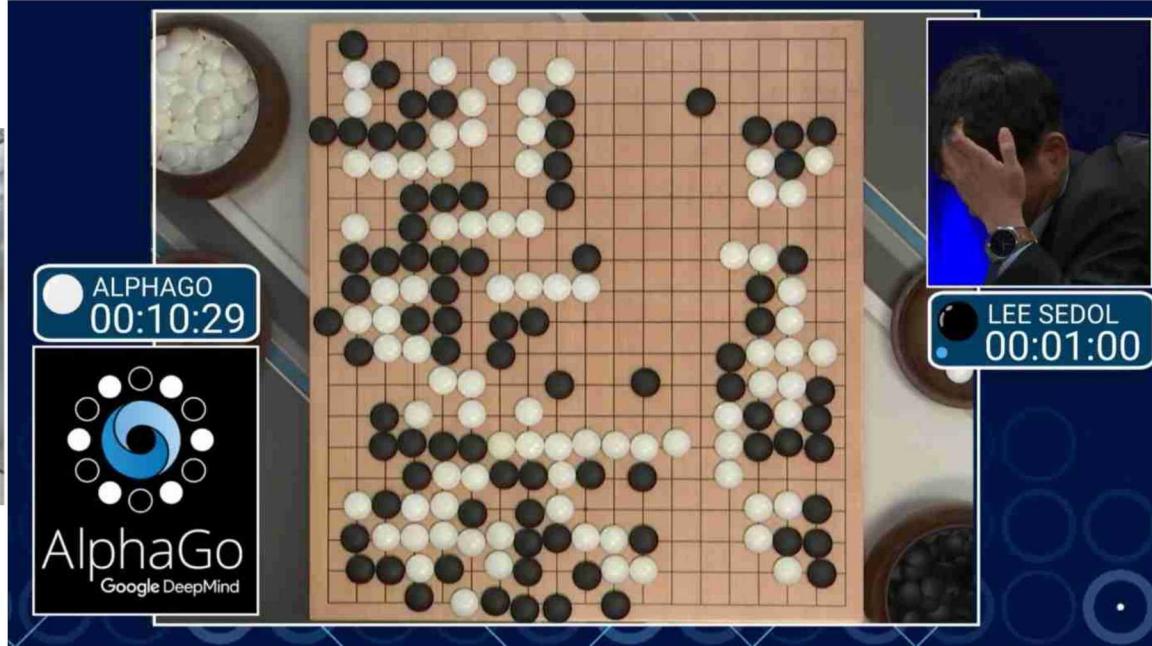
Low power  
High parallelism  
Simultaneous logic and memory



# AlphaGo (digital) vs. Lee Sedol (analog)



64 GPU  
19 CPU  
4 TPU



the brain ~10 Hz



100 billion neurons  
100 trillion synapses

AlphaGO: Megawatts



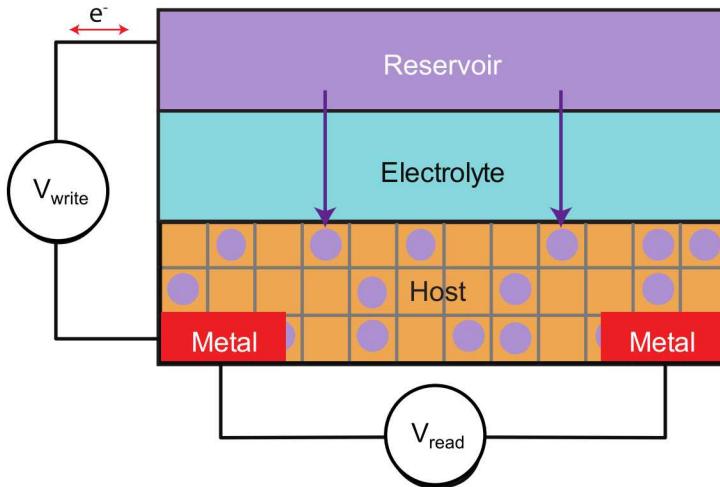
Human brain: 20 watt



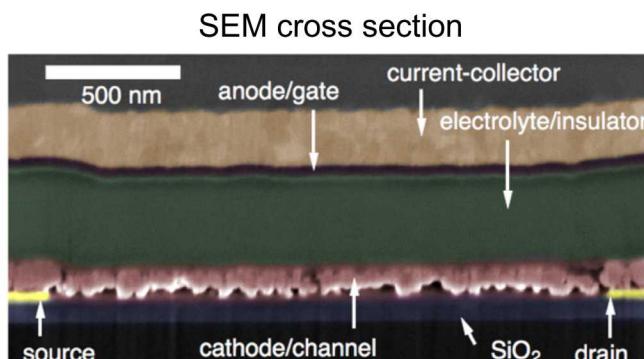


# Ion insertion, non-volatile redox transistor

## Non-volatile tuning of electronic conductance (weight)

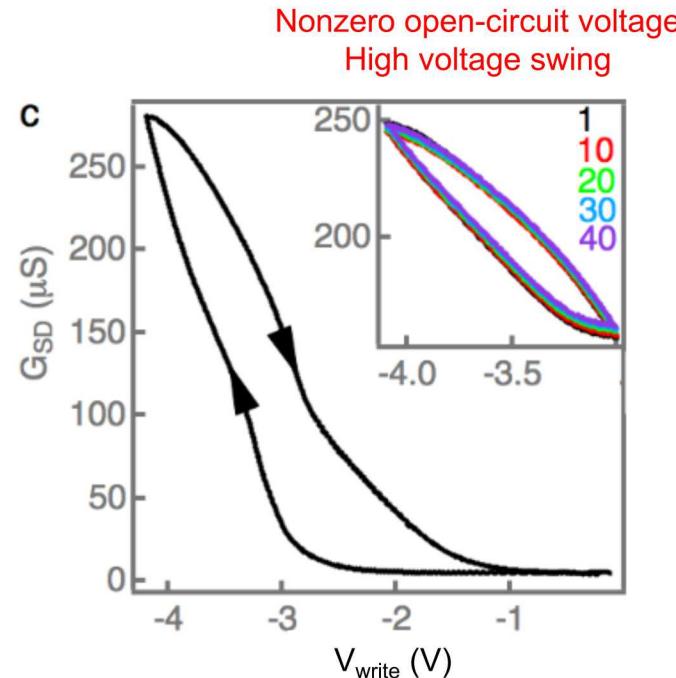


Modulating conductance via doping  
n-type:  $x Li^+ + xe^- + WO_3 \leftrightarrow Li_x WO_3$   
p-type:  $LiCoO_2 \leftrightarrow Li_{1-x}CoO_2 + xLi^+ + xe^-$



channel electrode:  $Li_xCoO_2$   
gate electrode:  $Li_xSi$   
solid electrolyte: LiPON

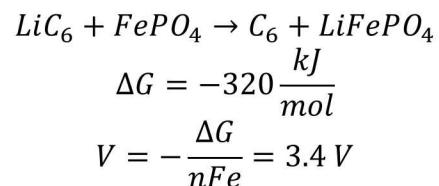
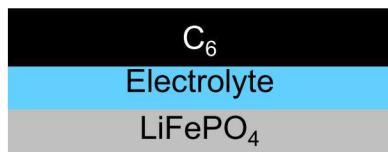
Fuller et. al., *Adv. Materials*, 29, 1604310 (2017)



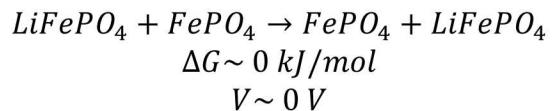
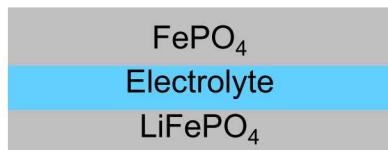


# Zero-volt ion insertion transistor

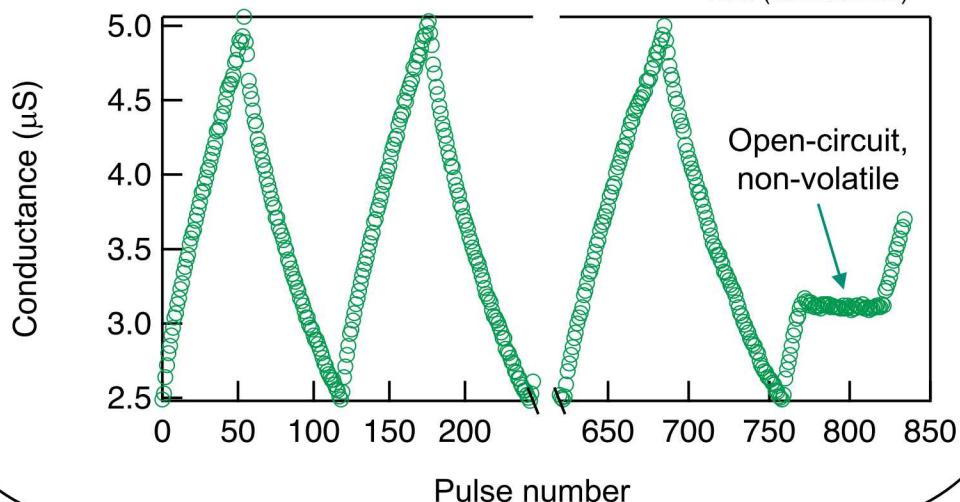
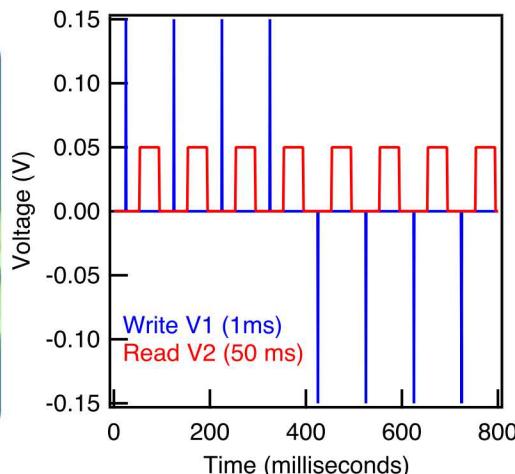
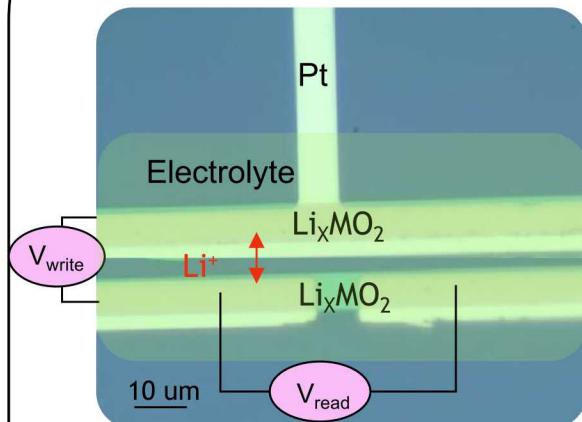
## Standard battery: Different lithium hosts



## Symmetric battery: Same lithium hosts



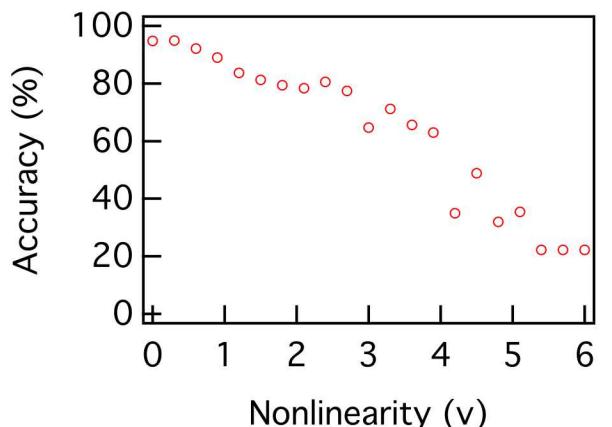
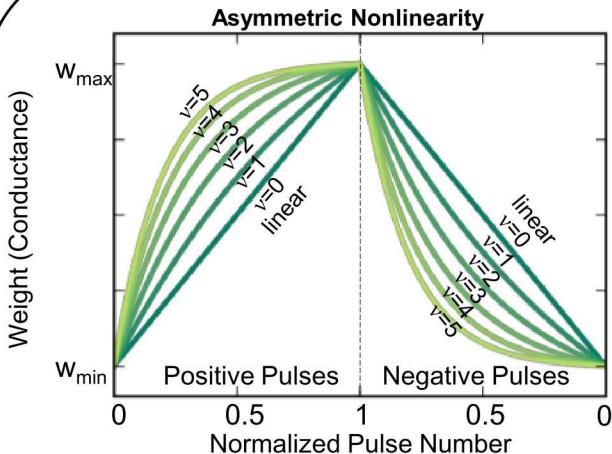
## Symmetric battery = zero-voltage transistor





# Ion insertion offers high linearity at low write voltages

Linear response allows more accurate and predictable updating of the weights



S. Agarwal et al. IJCNN, 2016

$$\begin{bmatrix} y_1 \\ \vdots \\ y_m \end{bmatrix} = \begin{bmatrix} w_{1,1} & \cdots & w_{1,n} \\ \vdots & \ddots & \vdots \\ w_{m,1} & \cdots & w_{m,n} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

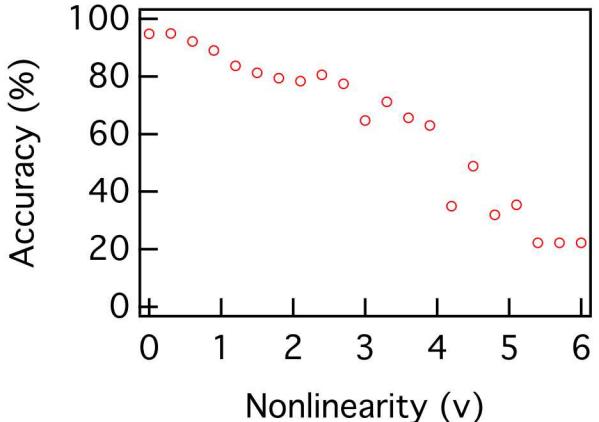
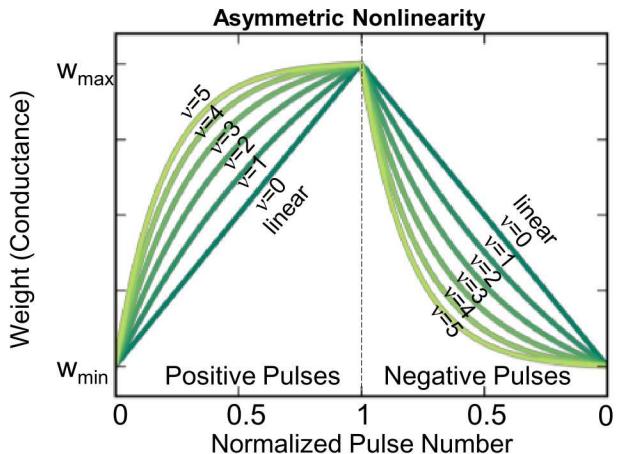


Linearity ensures each  $w$  can be tuned by a predictable amount.



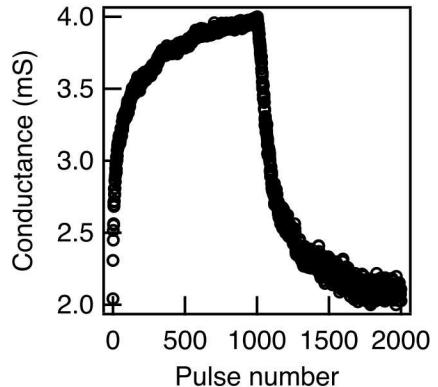
# Ion insertion offers high linearity at low write voltages

Linear response allows more accurate and predictable updating of the weights



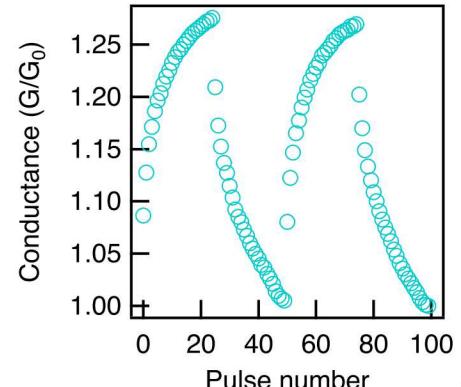
S. Agarwal et al. IJCNN, 2016

TaO<sub>x</sub> Memristor,  $\pm 1$  V  
 $v = 4.9$   
Accuracy:  $\sim 60\%$



R. Jacobs-Gedrim et al. ICRC, 2017

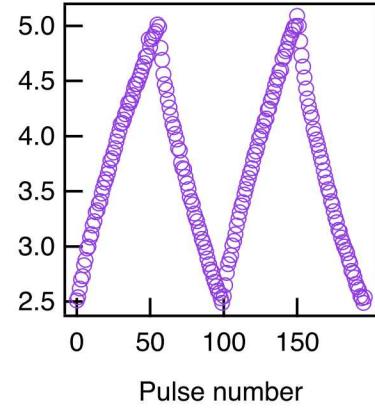
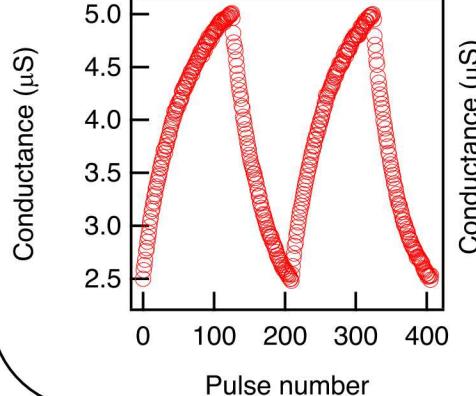
SONOS Flash,  $\pm 9$  V  
 $v = 3.7$   
Accuracy:  $\sim 70\%$



Ion insertion electrode

$\pm 50$  mV  
 $v = 2.27$

$\pm 200$  mV  
 $v = 0.67$   
Accuracy:  $\sim 85\%$

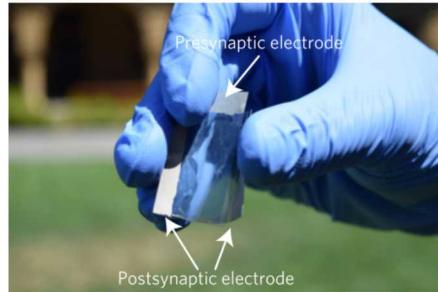
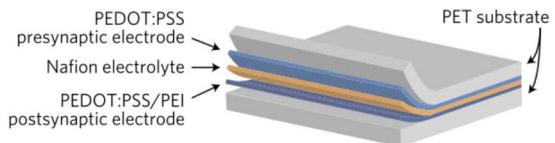
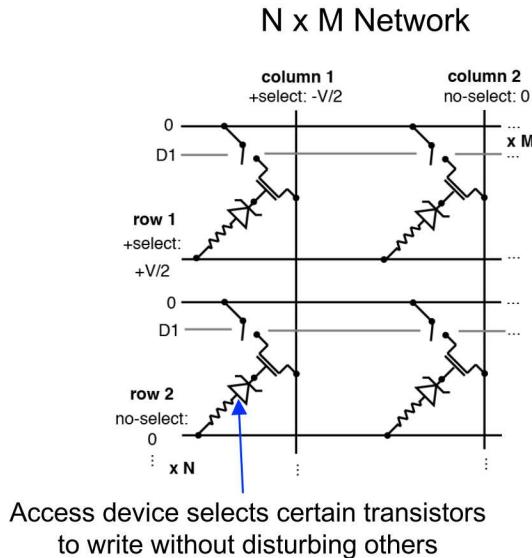


Ion insertion has much higher linearity and much lower power consumption

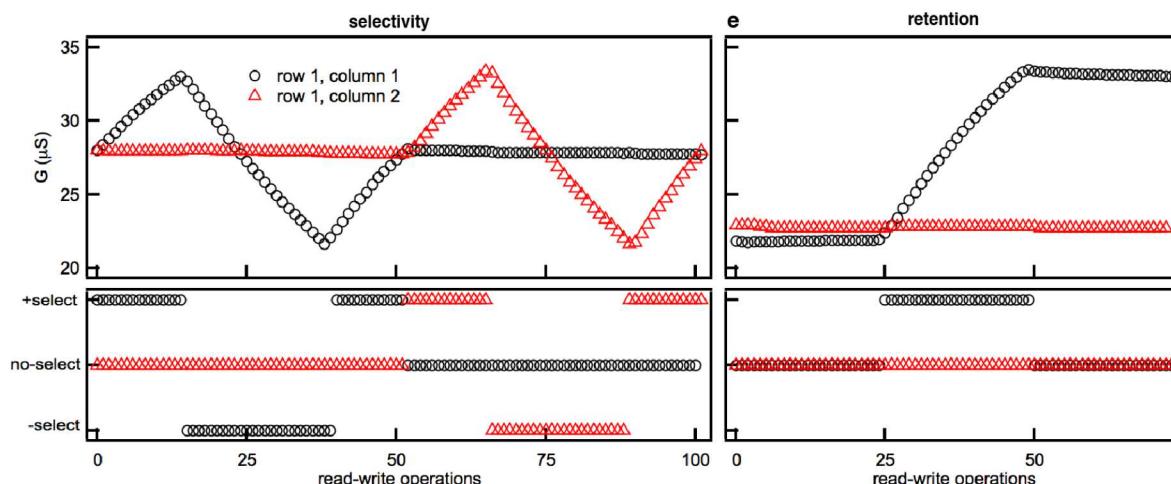


# From devices to networks

Can we build a parallel network of redox insertion transistors?



Polymer-based insertion transistors





# Acknowledgements

## Stanford

William Chueh

Jongwoo Lim

Will Gent

Norman Jin

Alberto Salleo

Scott Keene

## Berkeley Labs

Tolek Tyliszczak

David Shapiro

MIT

Martin Bazant

Dan Cogswell

## Sandia

Alec Talin

Elliot Fuller

Farid el Gabaly

Sapan Agarwal

Univ of Bath

Martin Bazant

Dan Cogswell

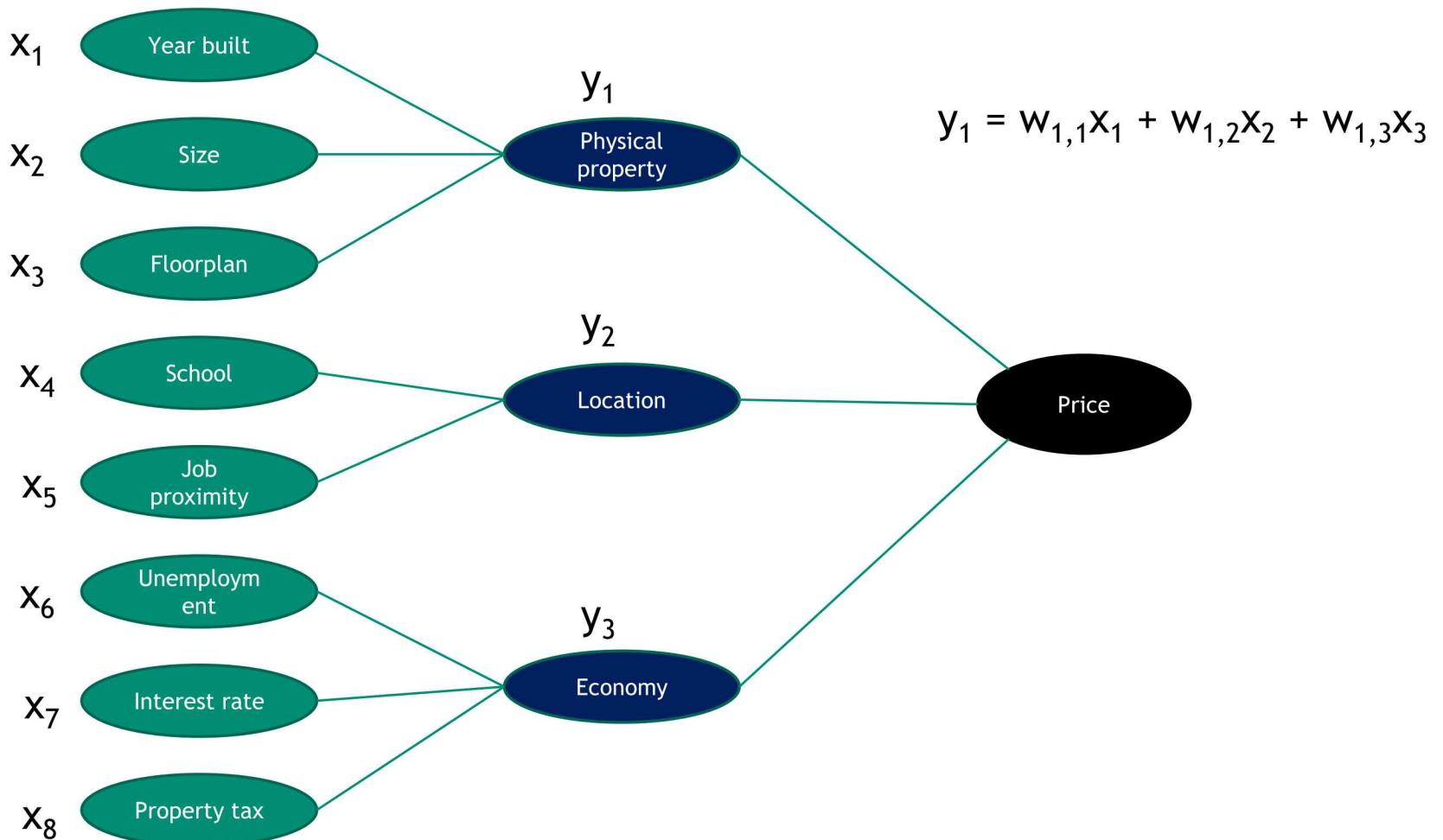


This work is supported by the Sandia National Laboratories Truman Fellowship Program, which is funded by the Laboratory Directed Research and Development (LDRD) Program. Sandia National Laboratories is a multi-mission laboratory managed and operated by National Technology and Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA-0003525.



# Matrix multiplication and artificial neural networks

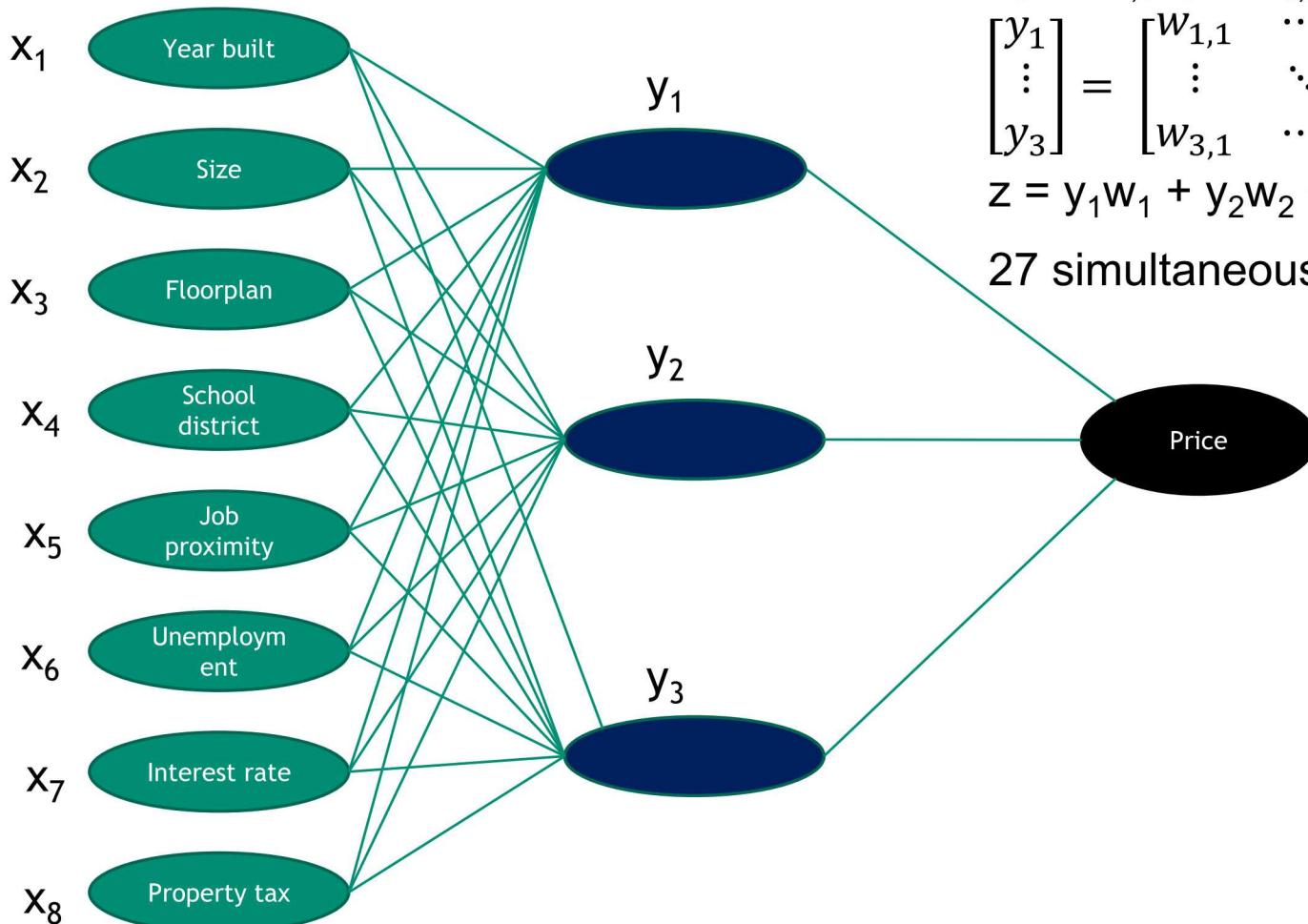
How much is my home worth?





# Matrix multiplication and artificial neural networks

How much is my home worth?



$$y_1 = w_{1,1}x_1 + w_{1,2}x_2 + \dots + w_{1,8}x_8$$

$$\begin{bmatrix} y_1 \\ \vdots \\ y_3 \end{bmatrix} = \begin{bmatrix} w_{1,1} & \dots & w_{1,8} \\ \vdots & \ddots & \vdots \\ w_{3,1} & \dots & w_{3,8} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_8 \end{bmatrix}$$

$$z = y_1w_1 + y_2w_2 + y_3w_3$$

27 simultaneous multiplication

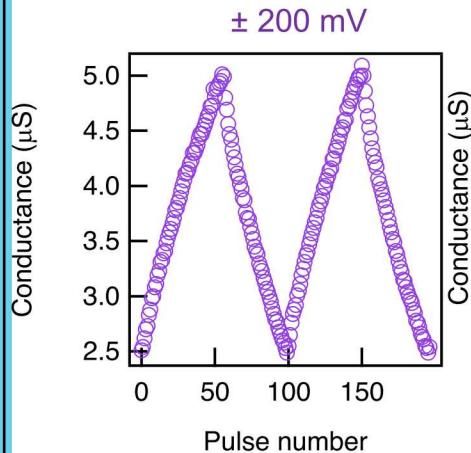
Neural networks often have 100 million weights

Google tensor processor unit does 65k simultaneous multiplications

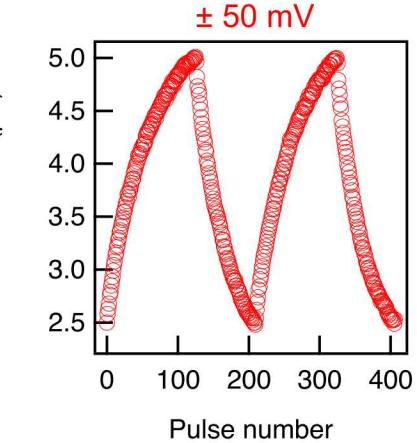


# Voltage pulse and linearity

Why does higher voltage pulses result in more linear response?



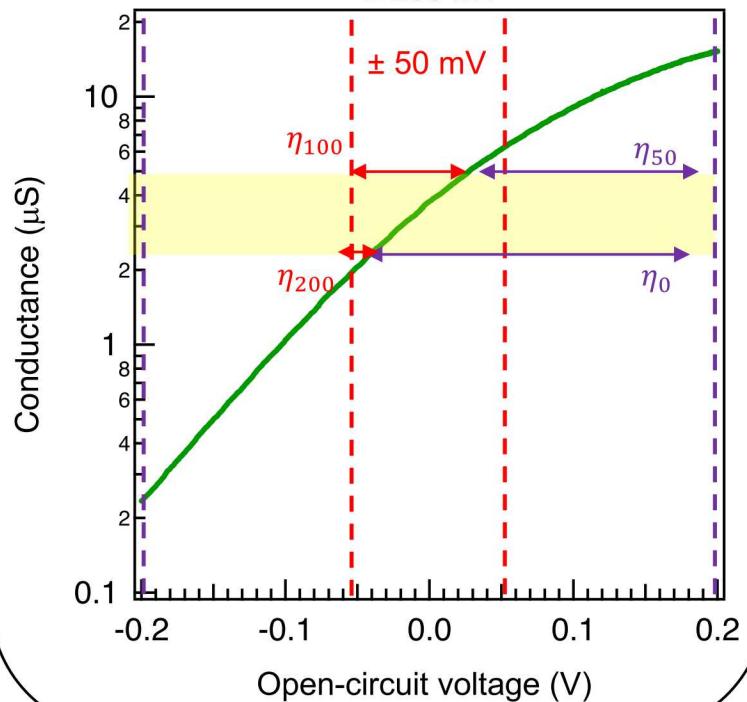
Overpotential decreases slightly with pulse number



Overpotential decreases drastically with pulse number

The open-circuit potential varies by  $\sim 100 \text{ mV}$  due to the entropy of lithium

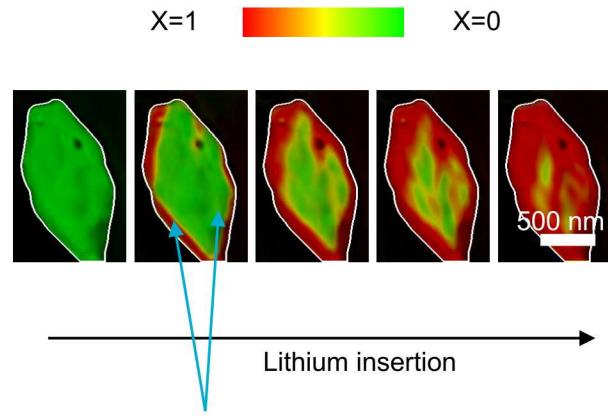
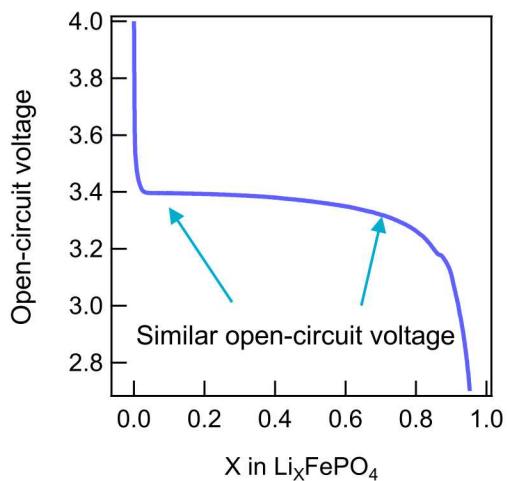
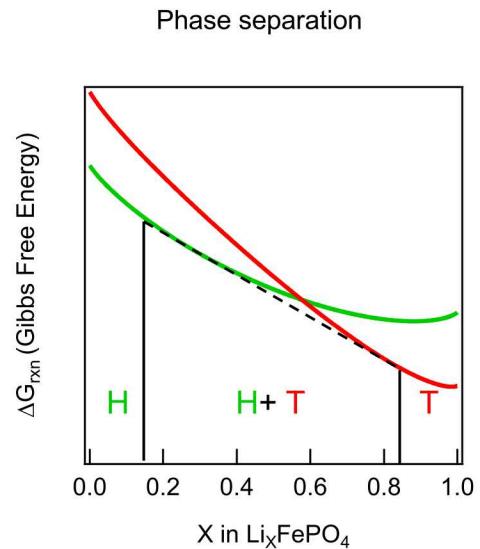
$$\mu = k_b T \ln \frac{x}{1-x}$$
$$\pm 200 \text{ mV}$$



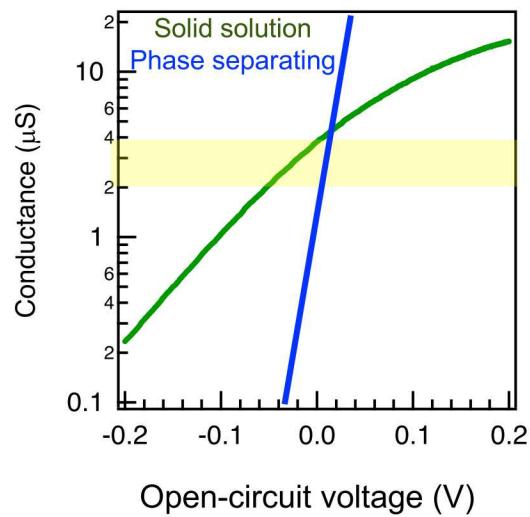
Goal: Find a material with higher  $dS/dV$  so the open-circuit voltage does not change with lithium?



# Phase-separation and voltage plateau



Different lithium doping concentrations  
Same chemical potential



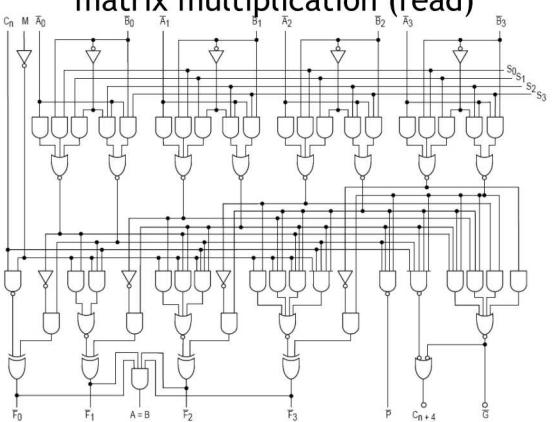


# Parallel analog processing

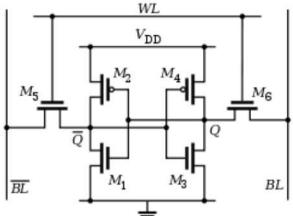
Aim: to mimic the brain and design a low-power device to conduct parallel analog matrix multiplication

## Digital logic

Arithmetic logic unit for matrix multiplication (read)

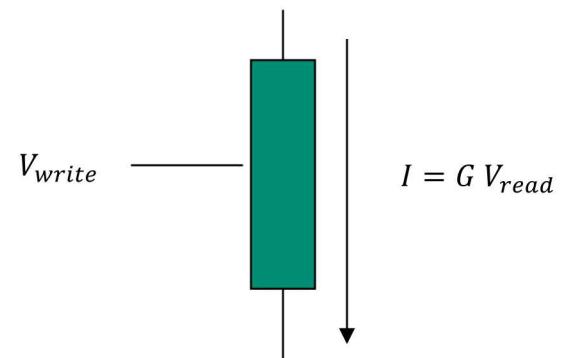


Static RAM for memory of the weights (write)



## Analog logic

Computation: modulate G continuously  
Memory: G should be non-volatile



$$\text{Power} = C V_{\text{write}}^2$$

Goal: Design devices that can linearly modulate the conductance with analog states, and retain that conductance (non-volatile)