

Energy Harvesting in Future IoT Devices

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University of Southampton and
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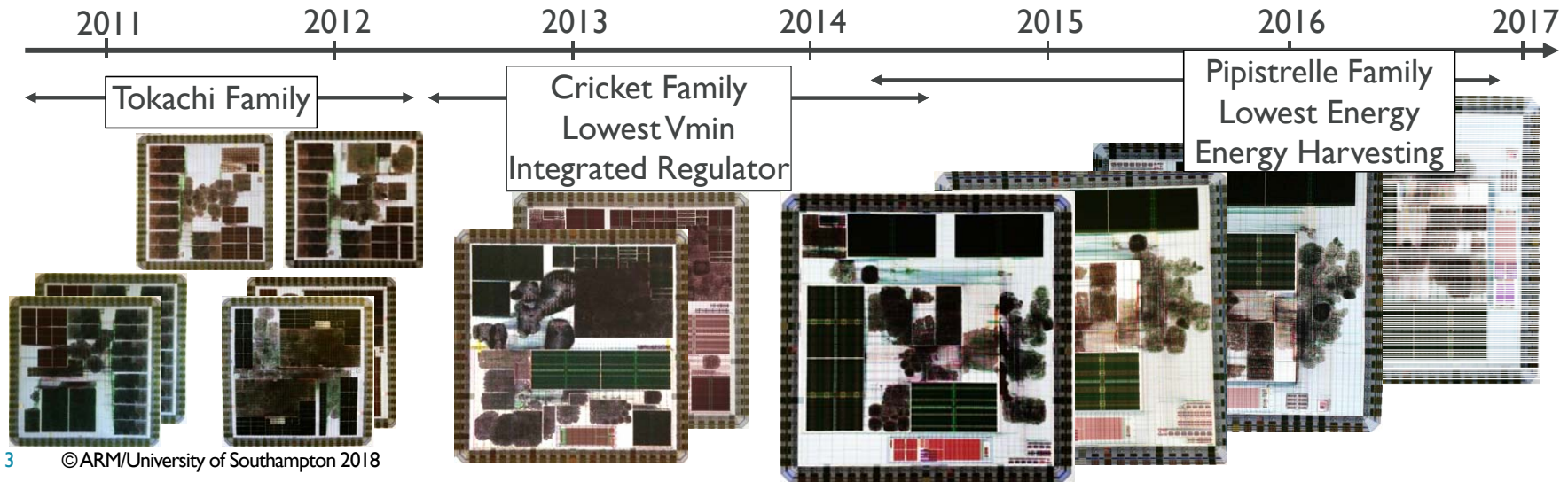
Arm and the University of Southampton

- Arm-ECS Research Centre (est. 2008)
 - Eight Southampton academics, Arm colleagues
 - Five jointly-supervised PhD students completed studies and employed at Arm (four more in the pipeline!)
- Centred around power-efficient computing
 - Mix of priority and 'blue-sky' research projects
 - Clock/power gating, energy-efficient IP watermarking, energy harvesting, run-time power modelling, system-level modelling and optimisation
 - 20+ jointly-authored publications
- Collaborative design and regular tape-outs
 - Ten research test chips taped-out over six years

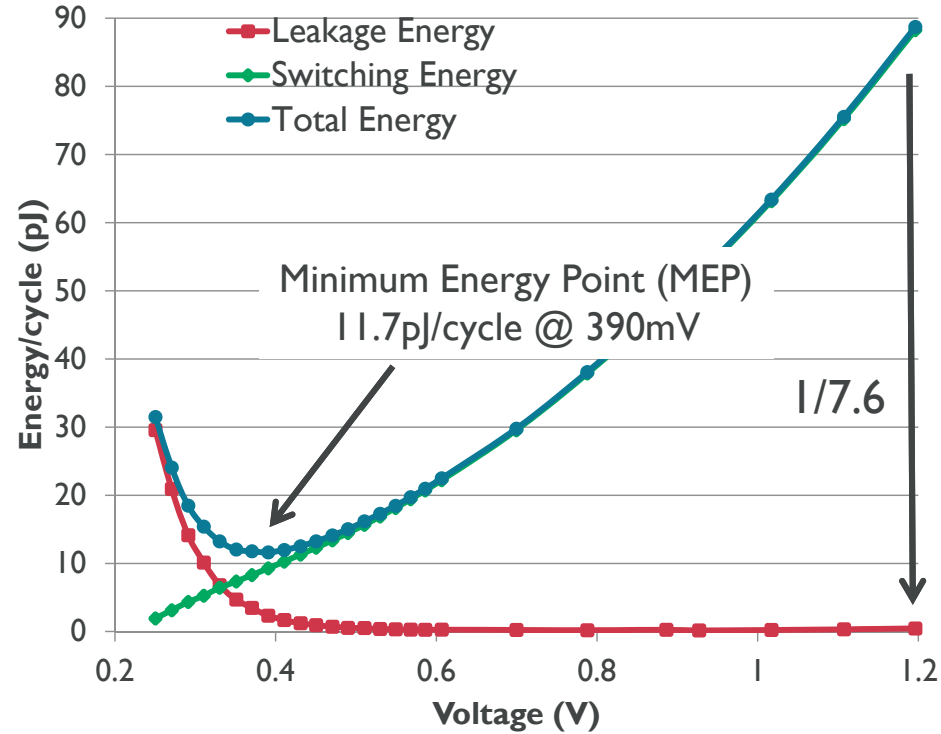
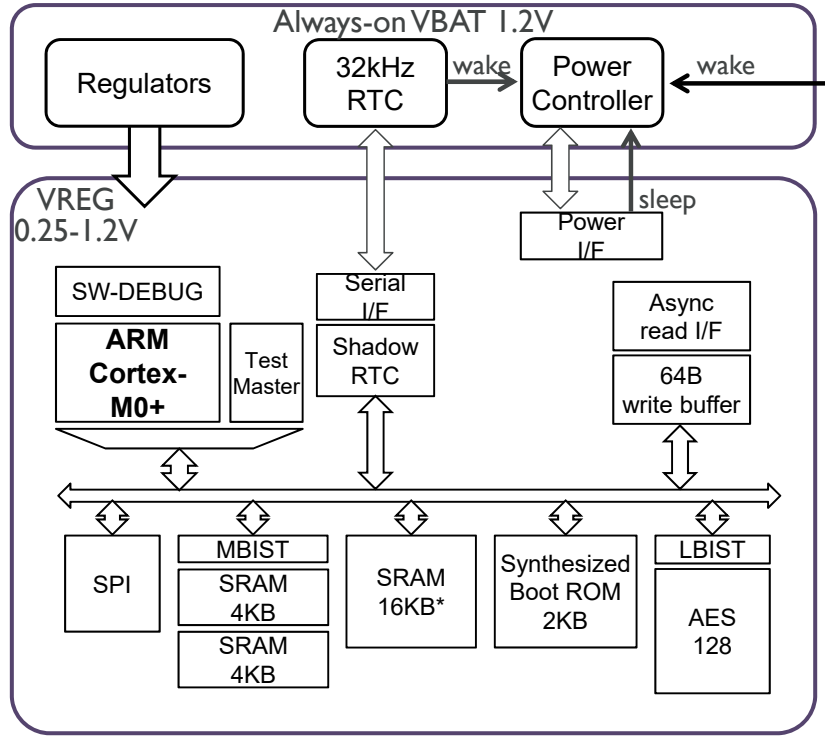


Collaborative Test Chip Program

- TSMC 65nm LP multi-project wafers through Europractice
- Not all projects need silicon – but many do benefit
- And being on a real tapeout is a great learning experience!



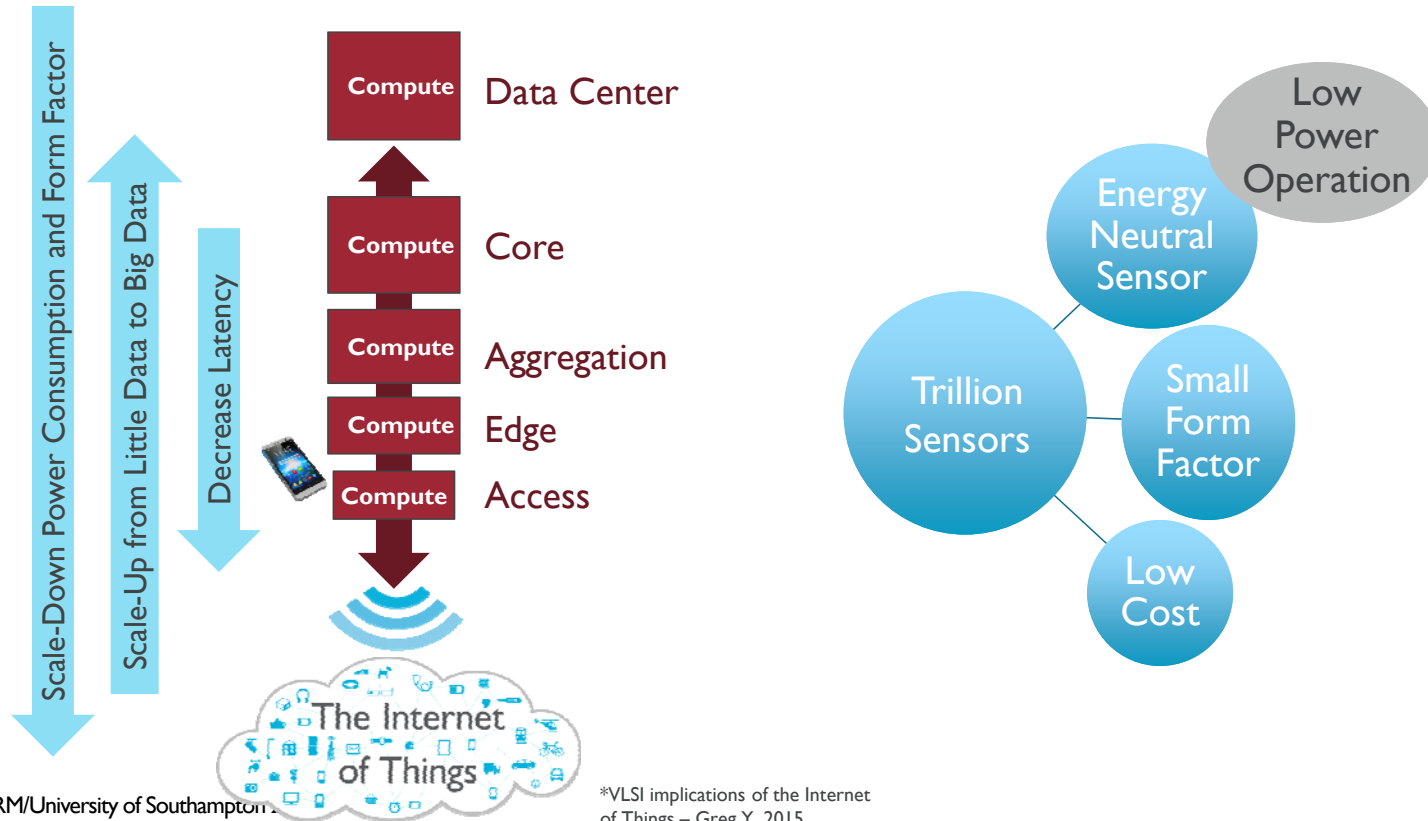
Cricket Sub-threshold MCU



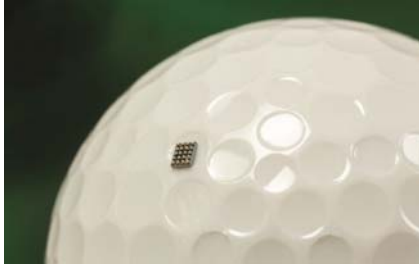
An 80nW Retention 11.7pJ/Cycle Active Subthreshold ARM Cortex-M0+ Subsystem in 65nm CMOS for WSN Applications, Myers et al, ISSCC 2015

Towards a Trillion Devices

Sensors and Compute in IoT



How small is small?



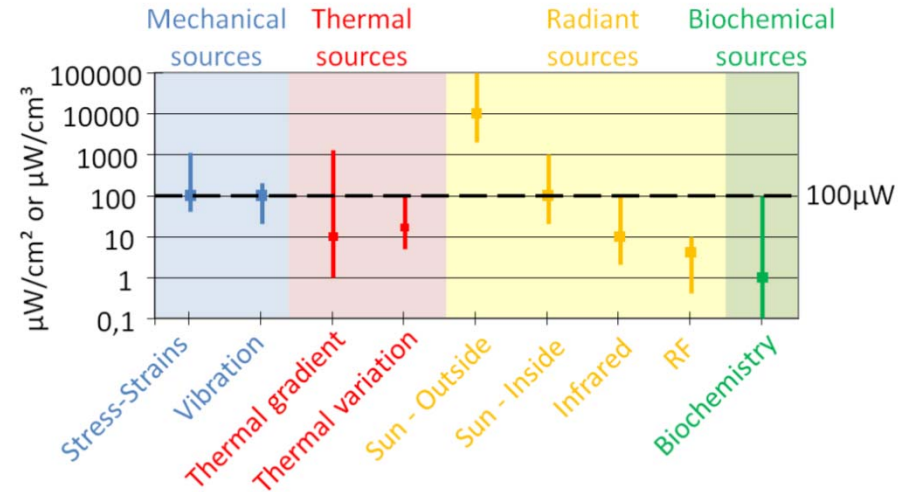
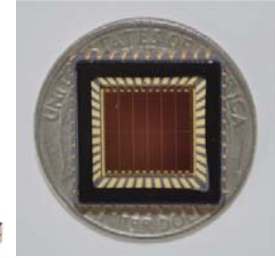
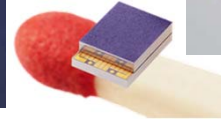
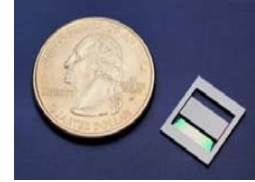
Freescale KL03

- Cortex-M0+
- 32KB Flash
- 2KB RAM
- 8KB ROM
- 12-bit ADC
- High speed comparator



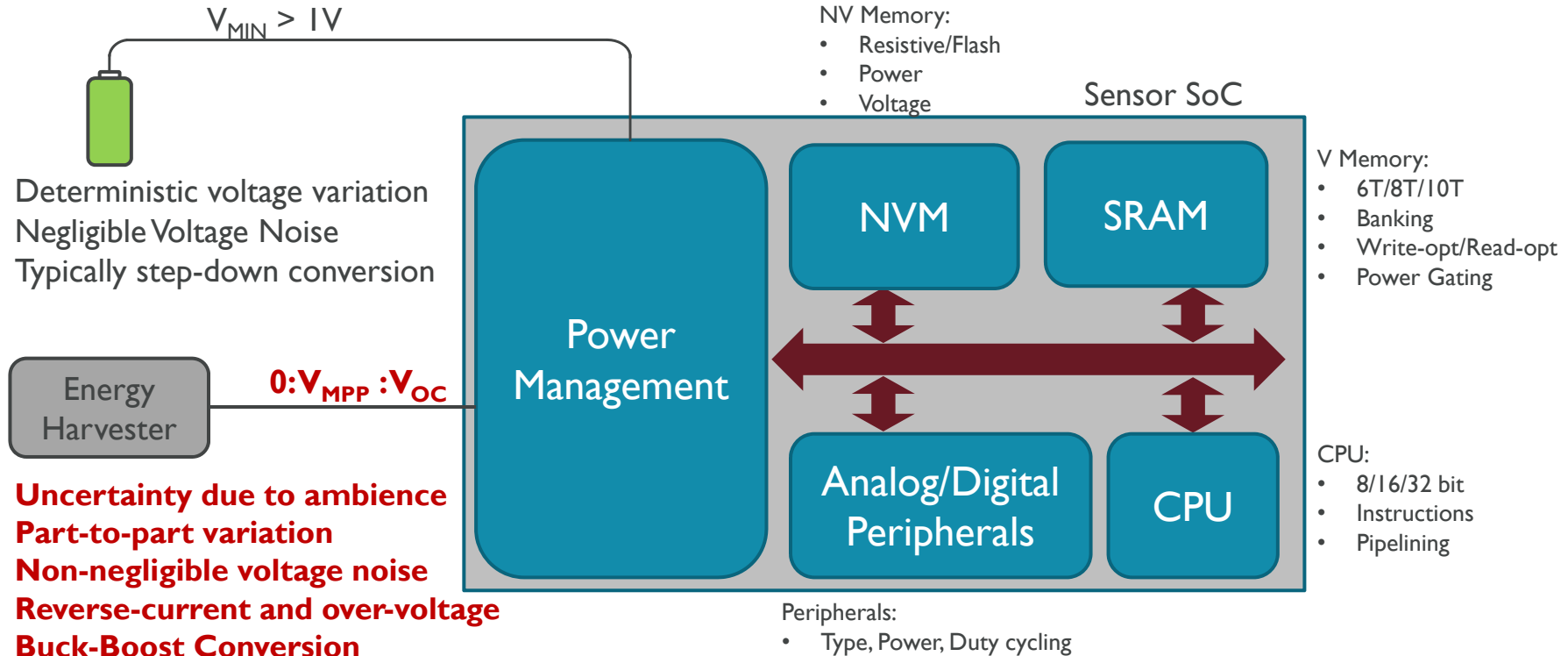
University of Michigan

- Cortex-M
- Custom memory
- Custom Radio
- Custom Battery
- Energy Harvesting



Variability in Energy Harvesting

Trillion Sensor Nodes: Designer's Perspective



Energy Harvesting in Everyday Life



Challenges of Energy Harvesting



Objectives for Energy Harvesting Systems

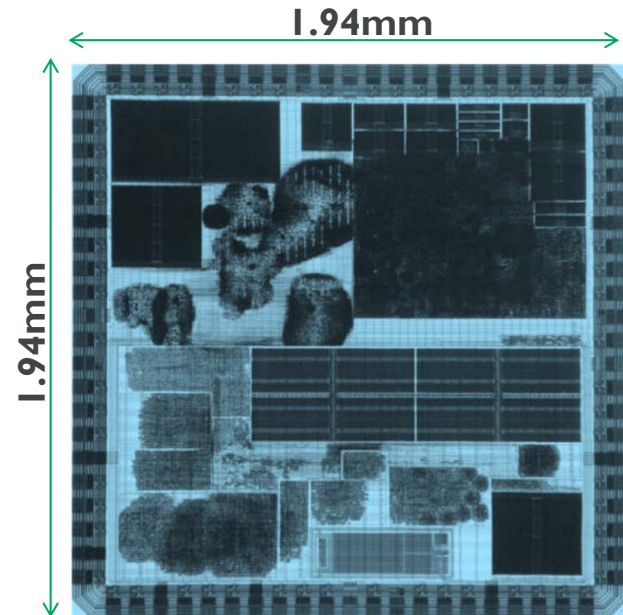
- Possible design objectives:
 - Works when I need it to!
 - Maintenance-free
 - Compact
 - Low-cost
 - Highly functional
 - Easy to deploy
 - Energy-aware
 - Able to cold-start
 - Able to work in most environments
 - Aesthetically pleasing
 - Wearable



How can we assess (and design for) variation in energy availability?

Coping with Variation

- Well-understood for conventional (battery-powered) systems
 - Process (fast, slow, typical)
 - Supply Voltage ($\pm 10\%$)
 - Temperature (0 to 85°C)
- Not defined for energy-harvesting systems
 - How to accommodate energy source variability? Spatial/temporal?
 - What are reasonable bounds for this?
- We *aim* for energy neutrality: harvesting at least as much as we use



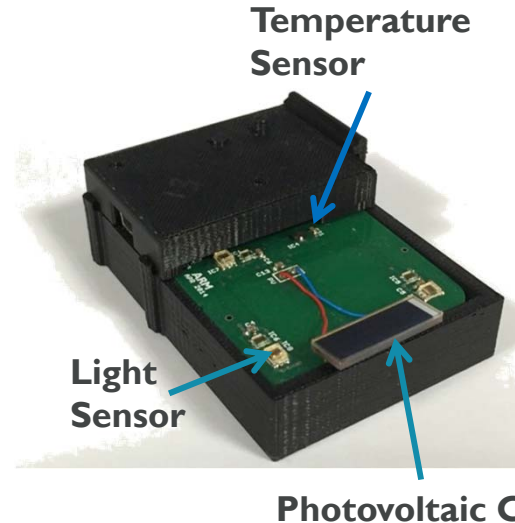
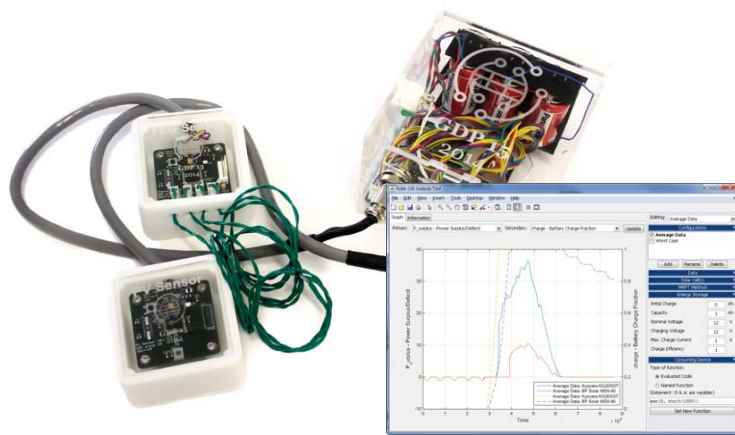
Coping with Variation

- **Temporal** variation
 - Use a larger energy buffer
 - Energy-aware behaviour: adjust duty-cycle, adapt to changing energy status
- **Spatial** variation
 - Balance sensing and routing tasks across network?
- Have to define these processes at design-time, within certain bounds
- How to estimate those bounds?



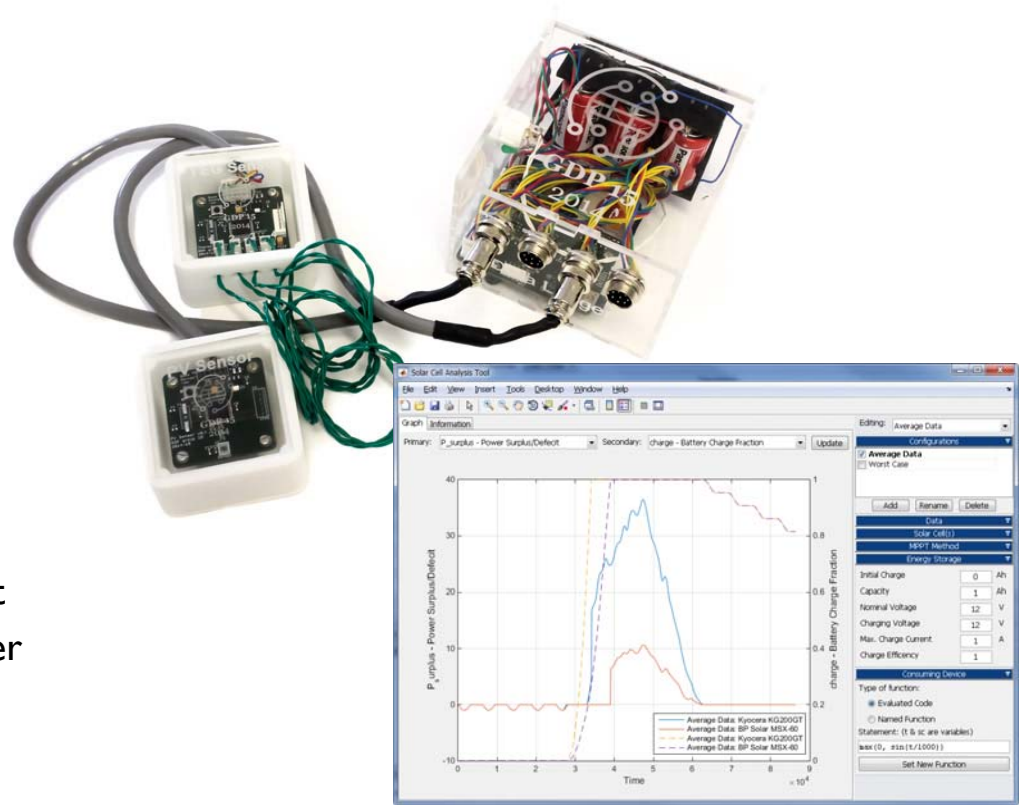
Ground Truth: Energy Availability Assessment

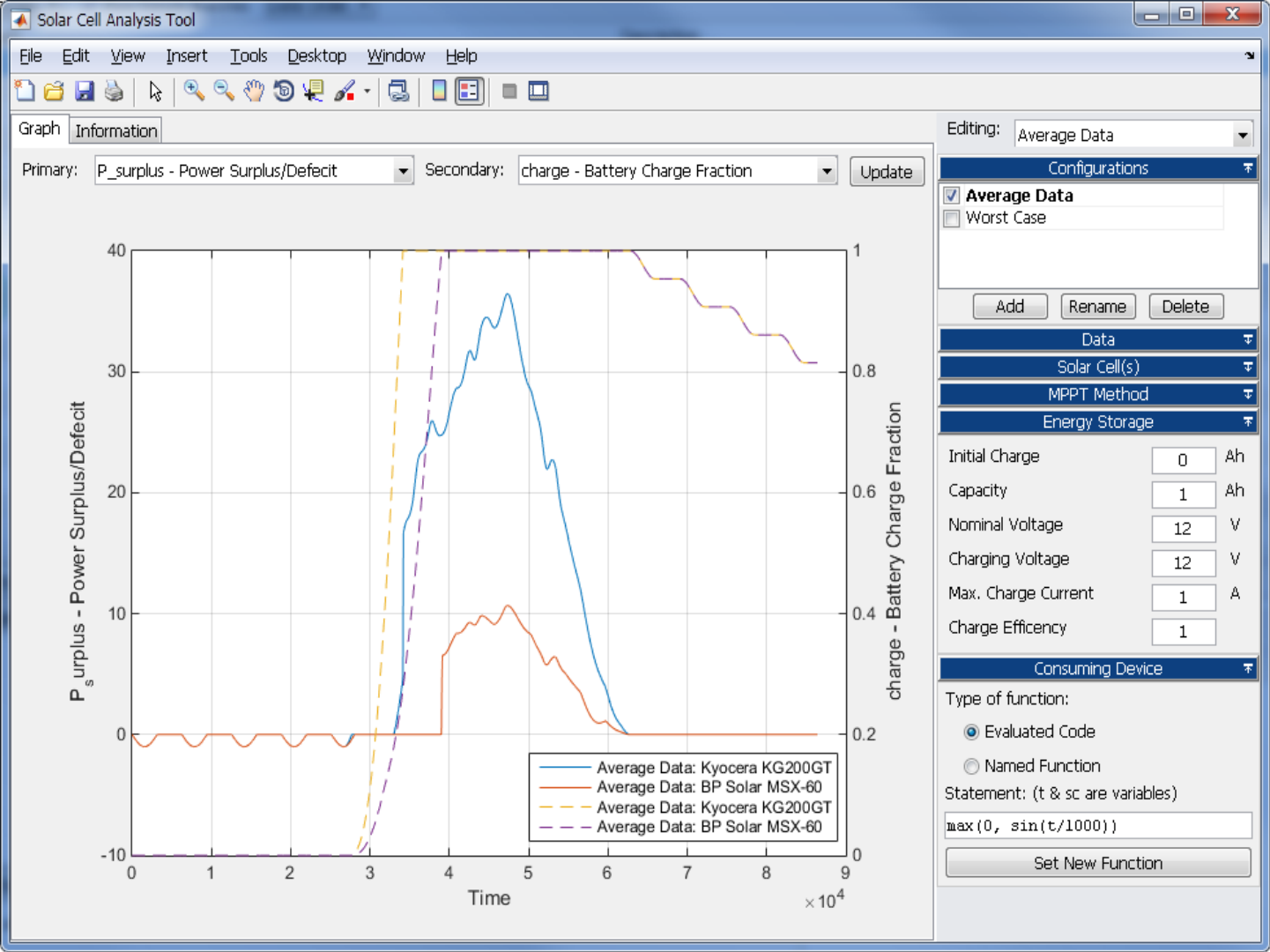
- How much energy is actually available?
- Enspect
 - A general tool, looks at environmental energy availability and models systems
- Micro Solar Evaluation
 - A specialised tool for micro PV cells, <1% error (current and voltage)

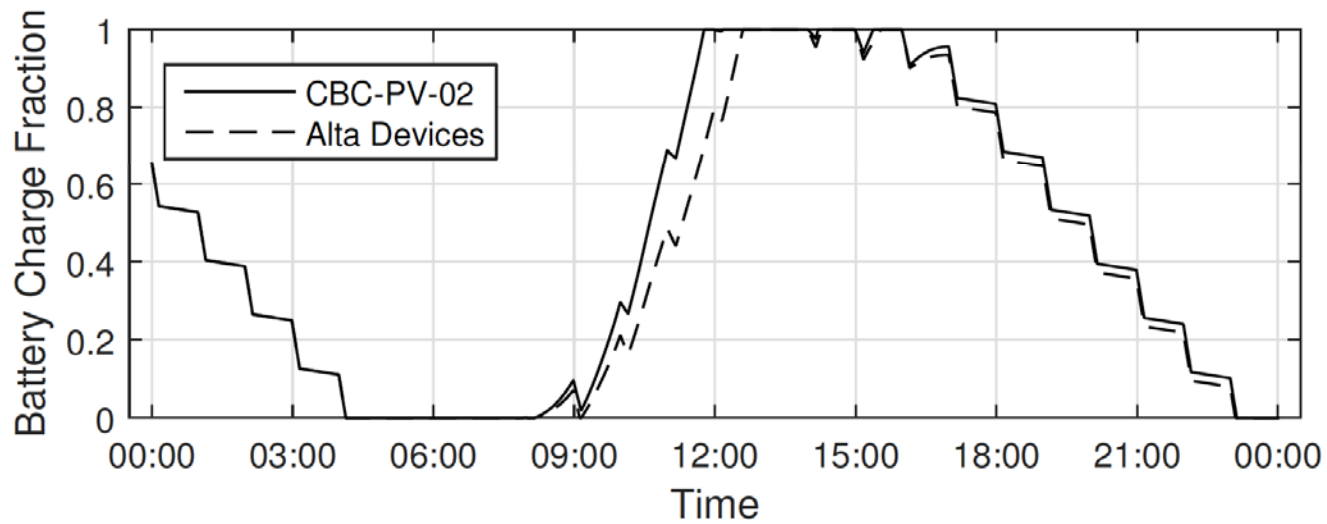
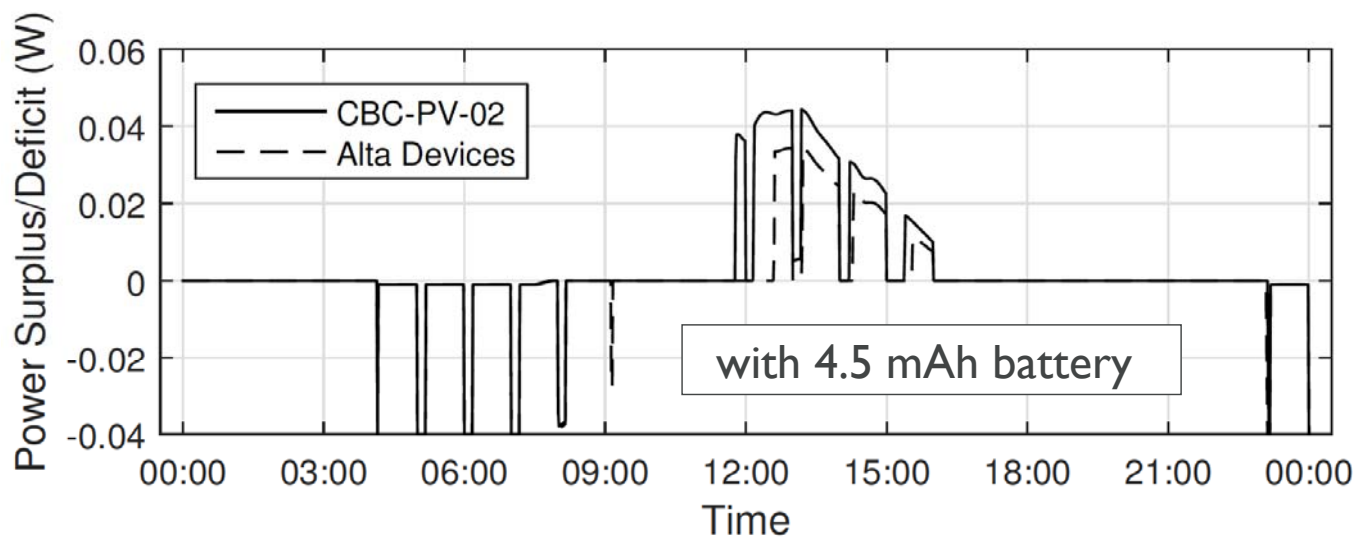


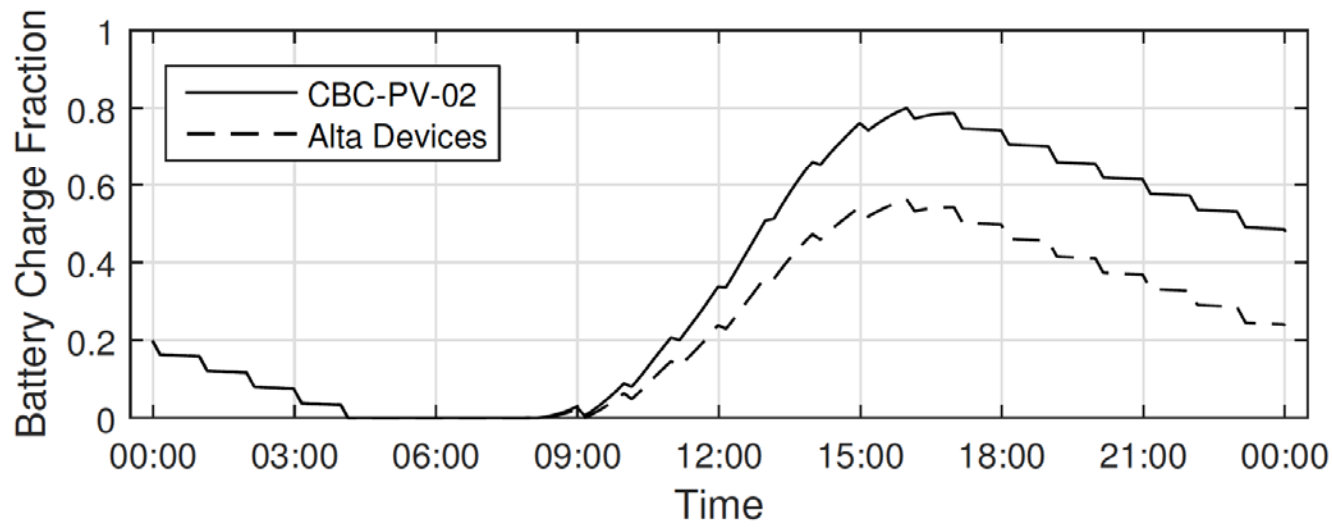
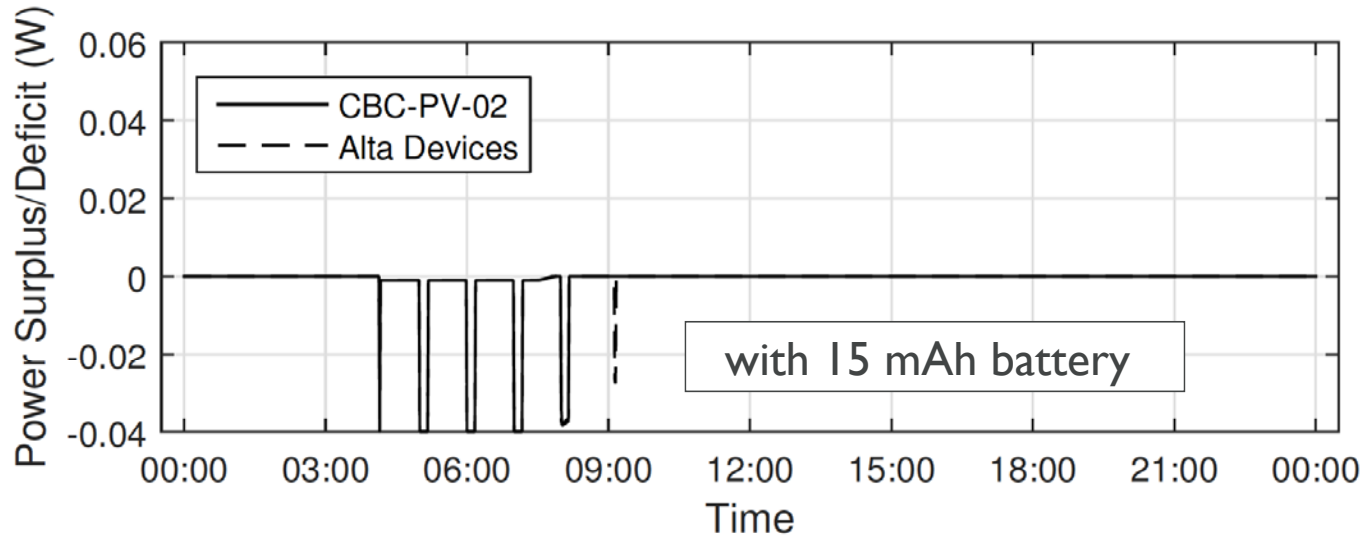
Enspect

- Logger and analysis software
- Models performance of harvester and rest of system
- Evaluate main design choices – flexible and fast
- New concept: energy *surplus* and *deficit* – assists with sizing
 - Storage: enlarge if in surplus and deficit
 - Harvester: enlarge if in deficit, but never in surplus
- Design recursively









Enspect

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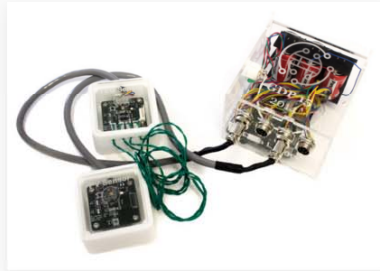
Enspect is a complete package for predicting the output of energy harvesting systems

Software last updated
6th Aug 2015

[Download](#)

Developed as a project of the School of Electronics and Computer Science at the [University of Southampton](#), Enspect aims to predict the power output of an energy harvesting system. It features a data collection unit which logs environmental data, and an analysis tool which processes this data to make predictions. Currently, an advanced photovoltaic cell model and simple thermoelectric generator model are provided.

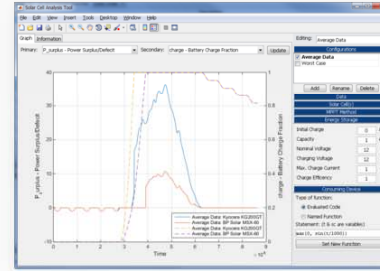
Data Collection Unit



Environmental data is collected via sensor modules collected to a central logger unit. Sensor modules were developed to allow **light irradiance modelling** for solar cells and **temperature sensors for thermoelectric generators**. Data is logged in CSV format to a standard SD card.

- Controlled **wirelessly** via Bluetooth.
- Long **battery life**.
- Modular **plug-and-play** design allows new sensors to be developed.

Analysis Tool



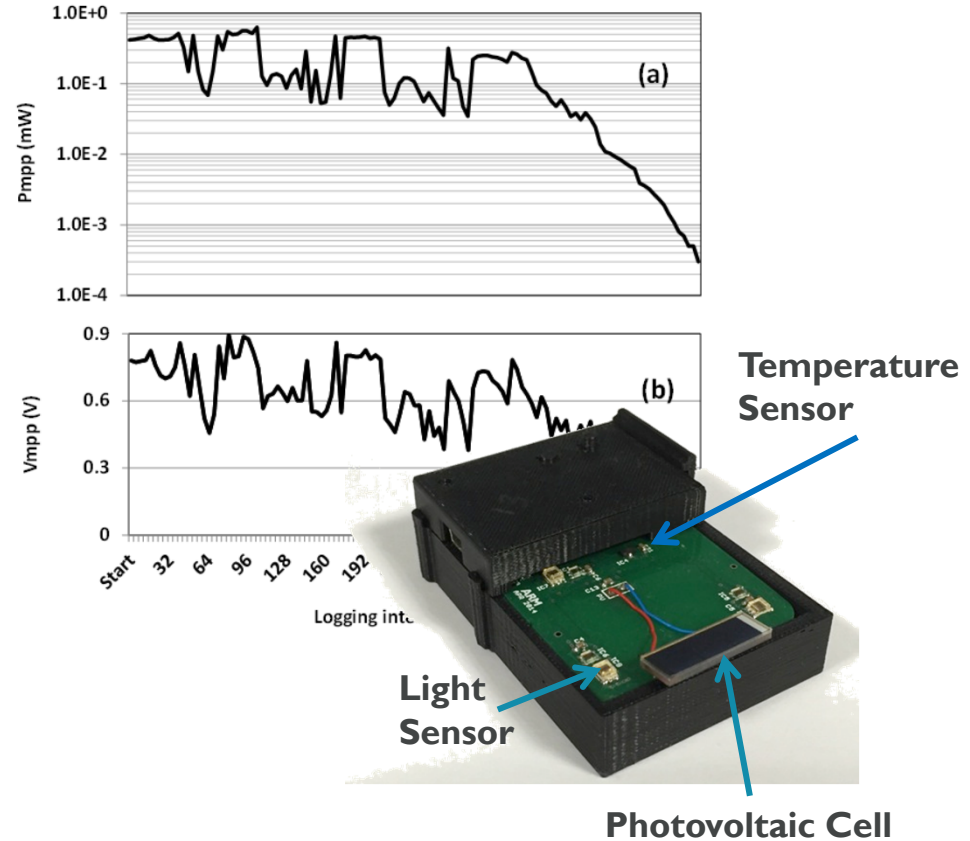
Complementing the data collection unit is an easy to use **graphical simulation tool**, able to predict power outputs from harvesters and take into account the power profile of the **consuming device** and any **battery storage**.

- Developed in **MATLAB**.
- Able to **compare** any number of system configurations.
- Models photovoltaic cells and thermoelectric generators.
- Gives a range of **graphical** and **tabular** outputs.

[More details](#)

Micro Solar Evaluation

- Specifically for evaluating cm^2 -scale cells in real environments (1 week+)
- Performs a full I-V sweep each time
- Better than 1% error
 - V measurement (0.1V...1.5V)
 - I measurement (10uA...5mA)
- Evaluate cell performance, design interface circuits inc. maximum power point tracking techniques



Using in System Design

- SPICE models of system components (inc. harvesters)
 - Variation-aware SPICE model
- Enables Monte-Carlo simulation of system performance

SPICE Description

```
.SUBCKT PVCELL P N lux=200
.param rp_scale = lux>1800?2:1
.param iph='-9e-12*lux*lux+70e-9*lux-5e-6'
.param is='5e9*pwr(lux,-9.5)'
.param is_var=agauss(0,1e-11,3)
.
.
IP N Pint 'iph'
DMAIN Pint N dpv
.
.
.MODEL dpv D (LEVEL=1 ...)
.ends PVCELL
```

Test Bench

```
RLOAD VP 0 'rload'
X1 Vp 0 PVCELL lux='illum'

.tran 1n 3u sweep monte=30 $$data=iload

.data illum_data
+illum 500 1000 1500 2000 2500
.enddata
```

Extract Data from
characterisation system



Parse measured data for
'remarkable points'



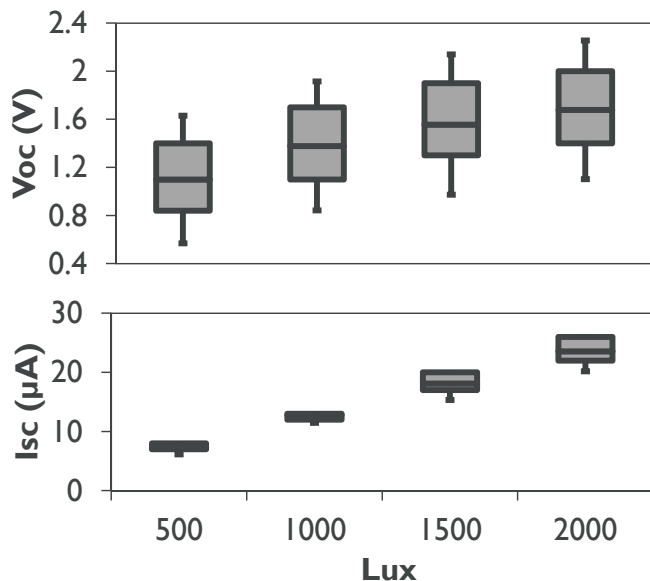
Derive fitting functions for
each parameter in model



Generate SPICE model
using fitted functions

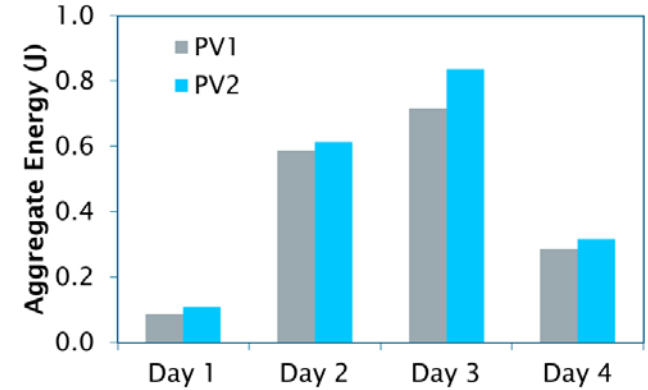
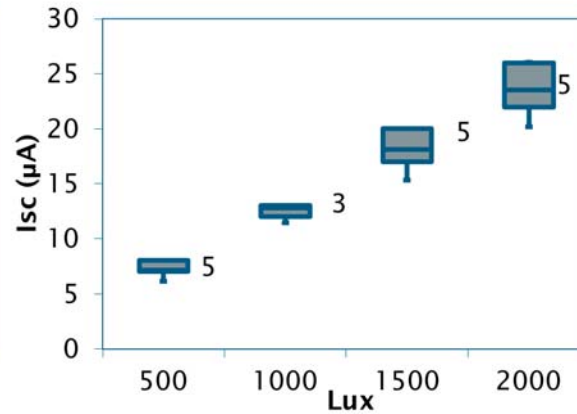
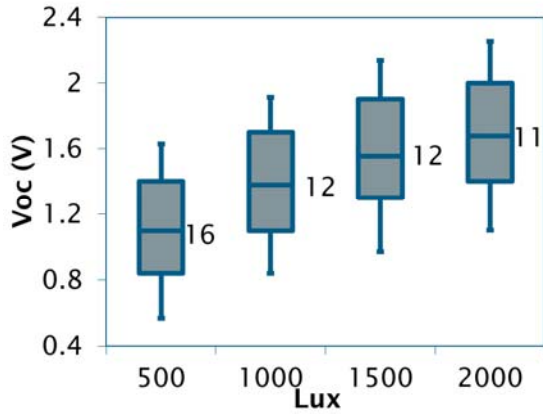
Micro Solar: Unexpected Findings

- Variation between cm^2 PV cells
 - Same manufacturer, same batch!

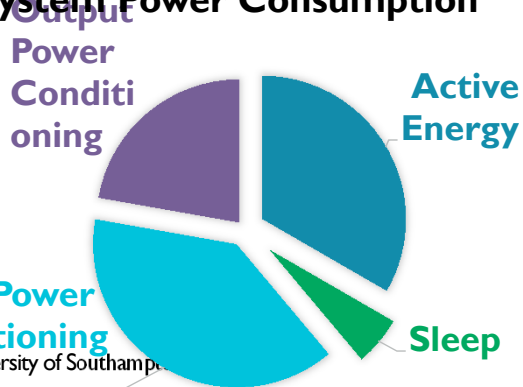


- Datasheets don't show this!
 - Typically state V and I (for light level)
 - No idea of variation
- So part-part variation is very significant (and unexpected)
 - Due to variation in doping on small cells?

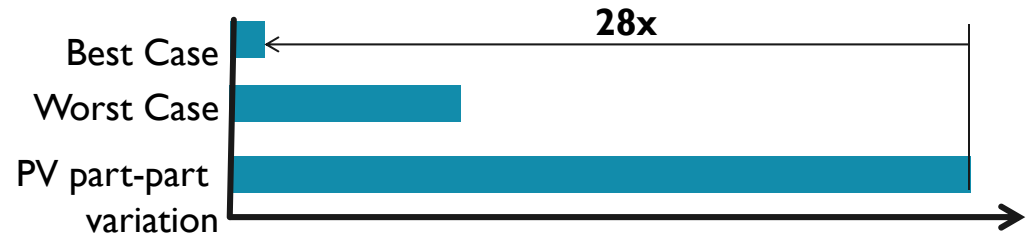
Modeling and Variability



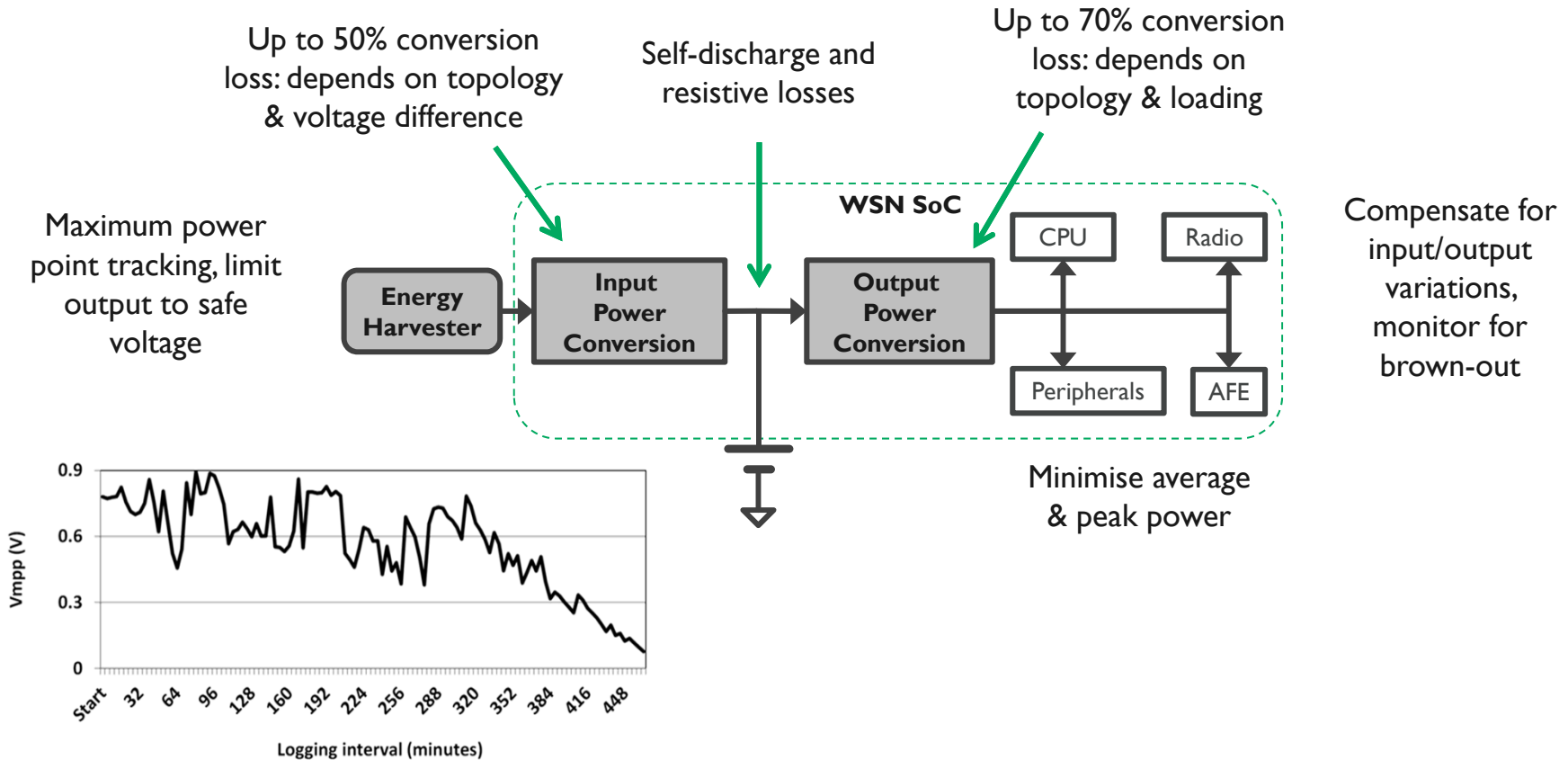
System Power Consumption



Required PV Area

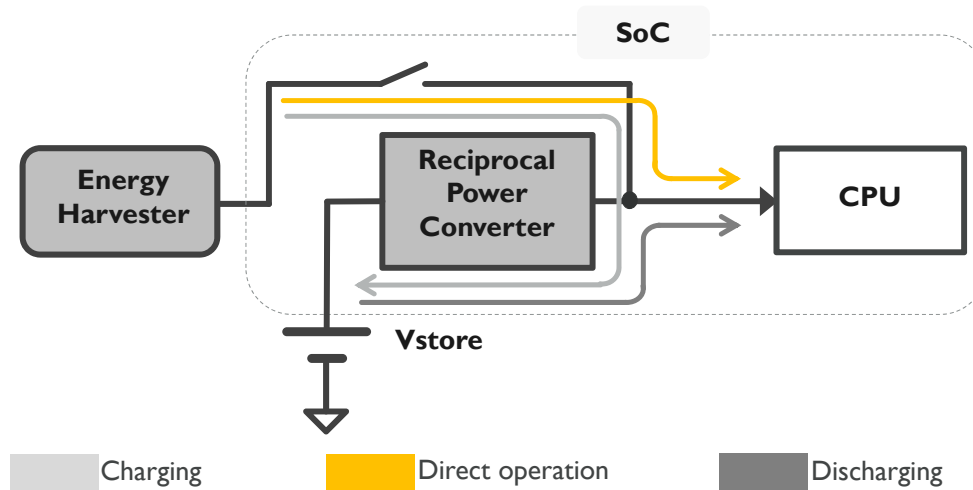


Voltage, Power and Energy



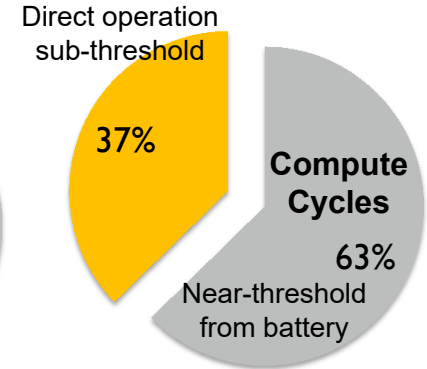
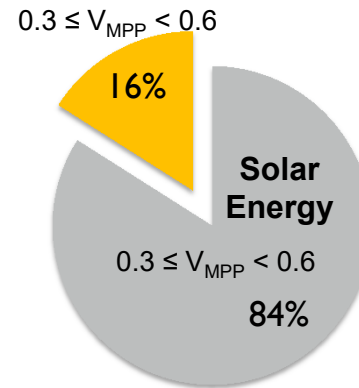
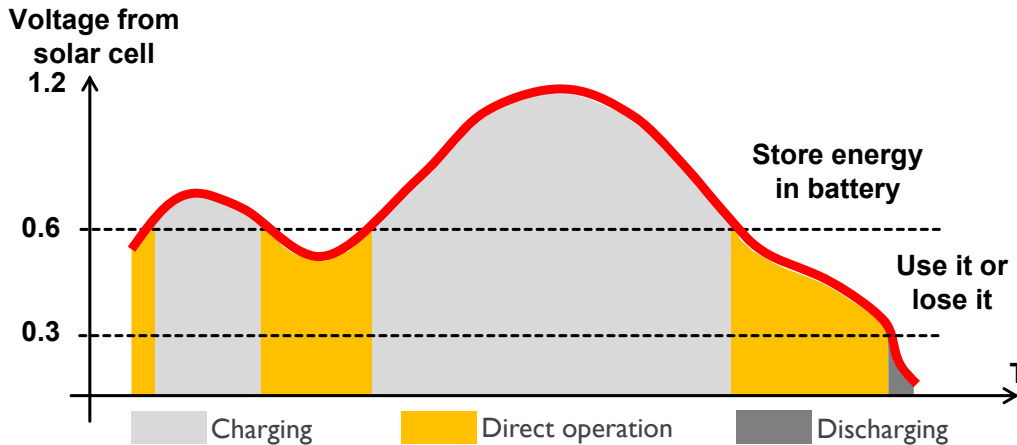
Efficient System Design

- Reciprocal converter, with direct operation
- Avoid the use of two converter stages
- Enables direct operation when conditions allow

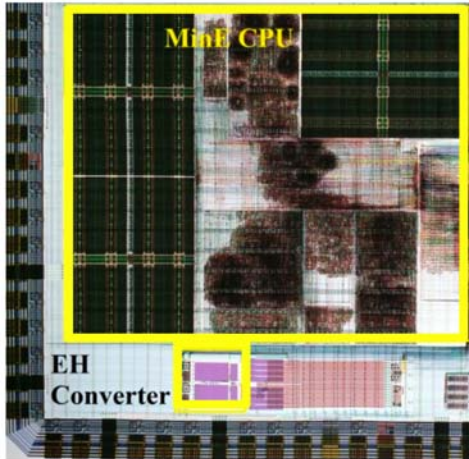
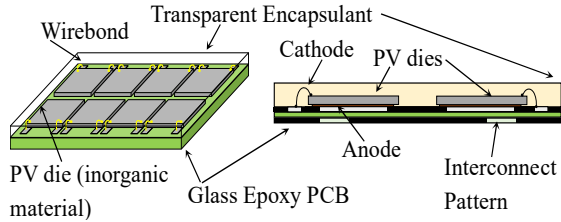


Energy Harvesting: Use it or Lose It!

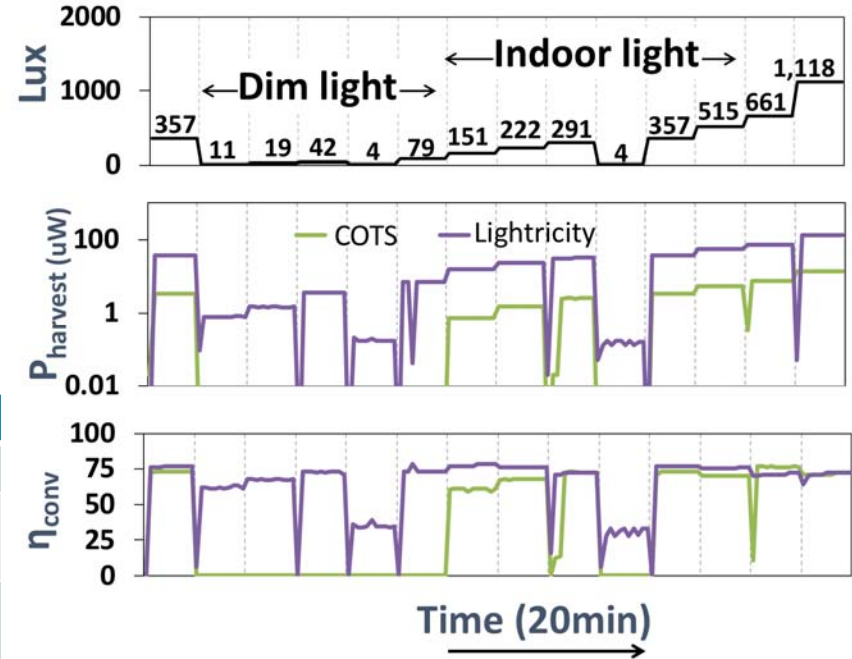
- Energy harvesters have variable output, causing waste of available energy
- Significant gains possible from direct-operation under some conditions



Solution: Co-Design

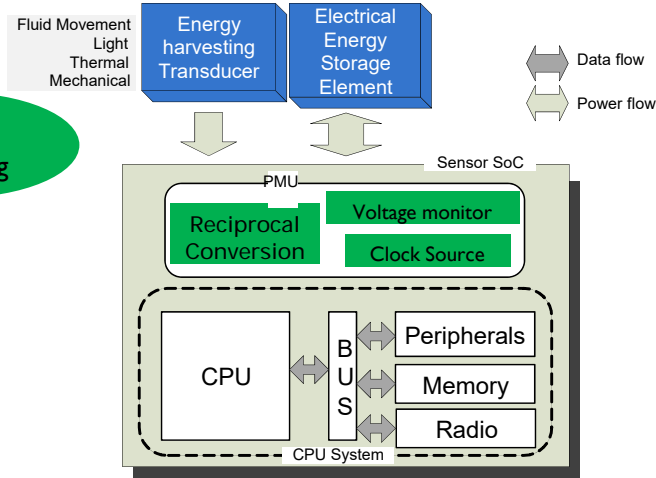


Block	Area (μm)
CHIP	1940 x 1940
LOGIC	1180 x 980
DCDC	175 X 210 (3%)

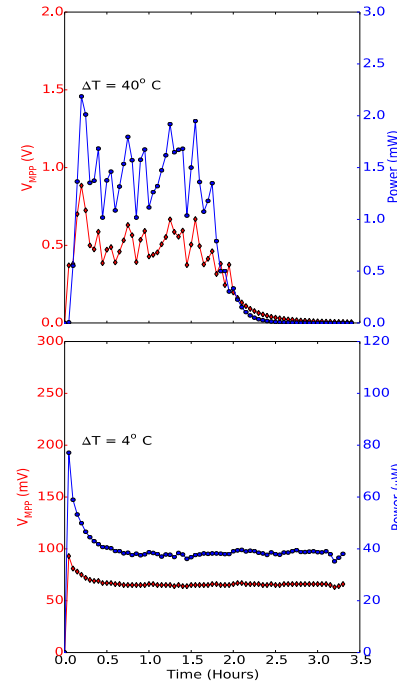


Solution: Holistic System Approach

PV cells modelling



[Savanth ISSCC'17, VSLID'17]



Energy-neutral system operation with 50 lux continuous light or 200lux for 2hrs/day , 20°C
 ΔT continuous or 50°C ΔT for $< 1 \text{ min/day}$

Conclusion

- Energy Harvesting and sensor system design for IoT
- Challenges with small scale energy harvesters
- Need for co-design, using real data, and holistic system solution
- Benefits from direct operation

- An open challenge: how to define *applications* to cope with variability

Acknowledgements:

- Arm: A Savanth, J Myers, P Prabhat, S Yang, A Kufel, D Flynn