

Modeling the Energy Consumption of Programs: Thermal Aspects and Energy/Frequency Convexity Rule

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Why focussing on energy saving for mobile computing?

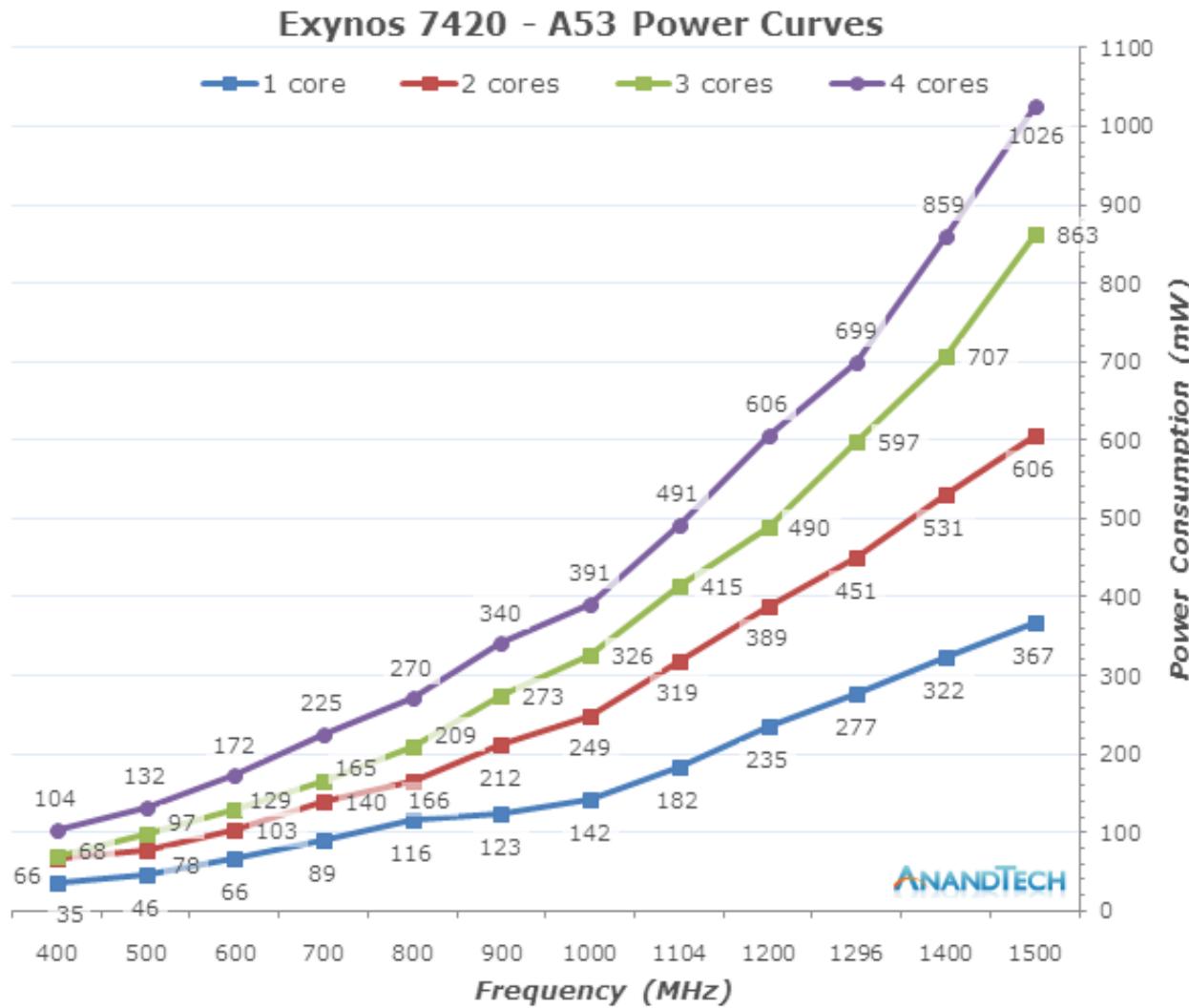
■ It is not *only* about the magnitude of saved energy:

- A smartphone CPU consumes between 60 to 500mW
- There were about 7×10^9 smartphones sold in the last 5 years, there will be 50×10^9 'smart objects' in 2022
- A worldwide saving of 30% would roughly mean about 280 MW for the smartphones, about 3 GW for the smart objects
- This would 'only' save between one tidal and one nuclear power station worldwide

■ Saving energy at the software level also is about a **natural-resource-free energy** saving

■ Focussing on mobile systems (e.g. a baystation on a drone): they are '*energy-critical*' : it is about being constantly looking for providing more autonomy with an unchanged QoS, with the same battery

The trend is moving towards providing more computational power



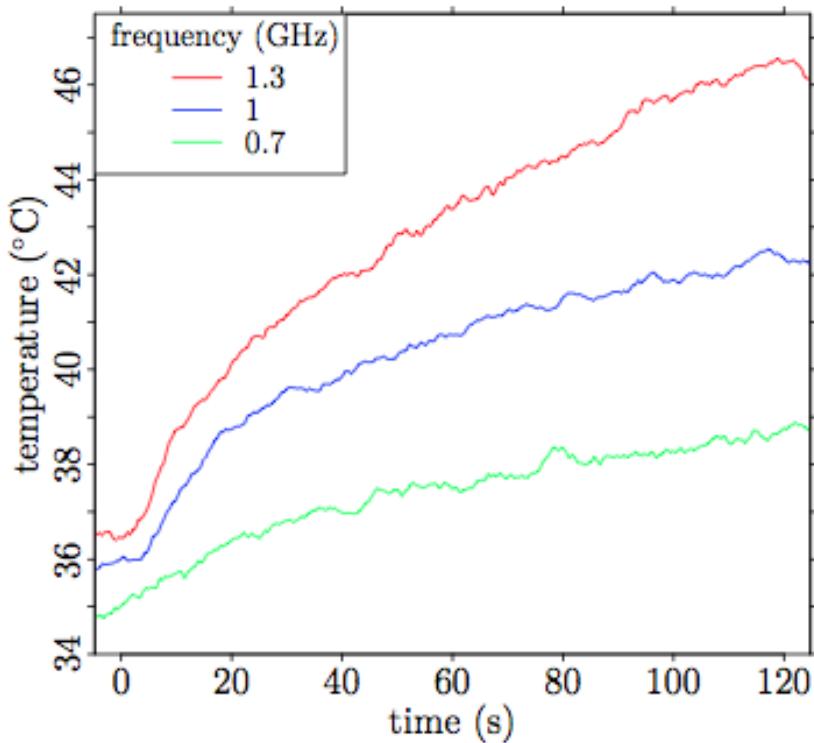


Thermal Behavior: Power-temperature rule Passive Cooling rule

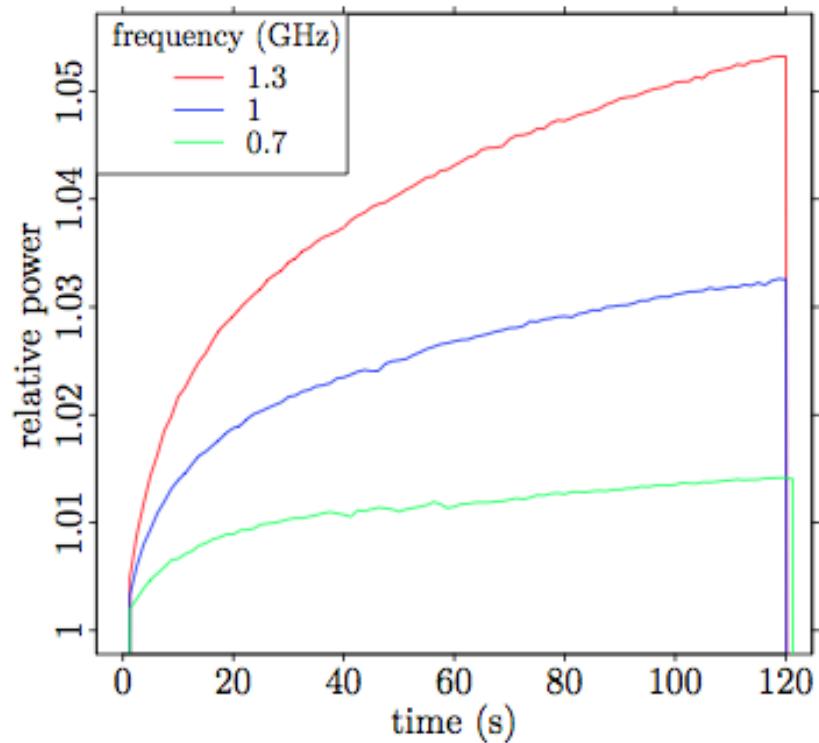
Temperature impacts energy consumption



$$P = a_1 e^{T/a_2} + a_0$$



(a) Temperature



(b) Power

An increase of 10% of temperature generates
an increase of 5% in power consumption

Small size mobiles have no fan Passive Cooling Rule

The passive lumped system's transient thermal behavior is defined by

$$mC \frac{dT}{dt} = \text{internal heat generation} + \text{radiation} + \text{convection}$$
$$mC \frac{dT}{dt} = P(T) + \epsilon\sigma(T^4 - T_a^4) + h(T - T_b),$$

which has the solution

$$t(T) = -\frac{1}{\kappa_4} \left(A \ln |T - \omega_1| + \frac{C}{2} \ln |(T - \alpha)^2 + \beta^2| \right. \\ \left. + B \ln |T - \omega_2| + \frac{\alpha C - D}{\beta} \arctan \left(\frac{T - \alpha}{\beta} \right) + c_o \right).$$

Approximations do exist for small areas



Contributions on thermal behavior

- Necessary for reproducible measurements and for accurate energy consumption models
- Power–temperature relationship
- Approximations for practical uses, particularly for online usages (embedded systems, radio mobiles,...)



EFCR: the energy – frequency convexity rule

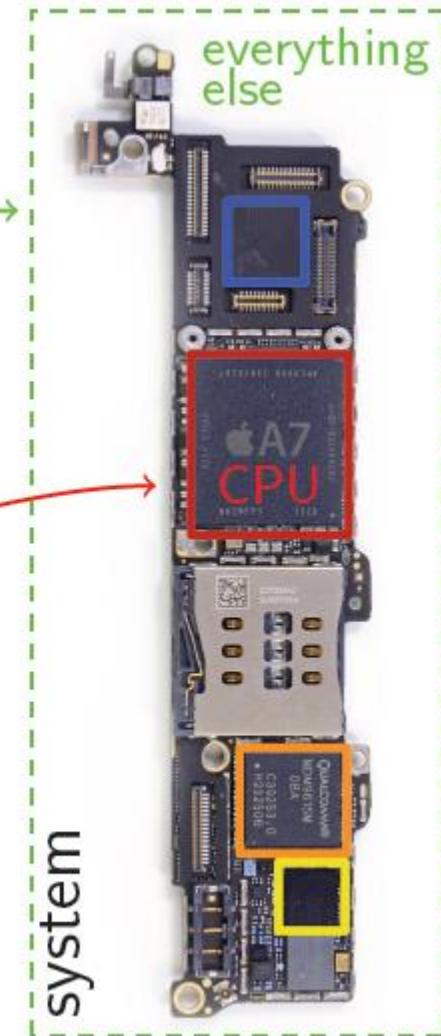
Fragmenting energy consumption per system module

- System's energy consumption E_{sys} definition

$$\begin{aligned} E_{\text{sys}} &= \int_0^{\Delta t} P_{\text{total}} dt \\ &= \int_0^{\Delta t} (P_{\text{cpu}} + P_{\text{back}}) dt; \end{aligned}$$

- Examples of P_{back} include:
 - ▶ LCD screen,
 - ▶ radio interface,
 - ▶ power supply;
- If P_{cpu} and P_{back} are sufficiently constant over time

$$E_{\text{sys}} = (P_{\text{cpu}} + P_{\text{back}}) \cdot \Delta t;$$





Power and time model

Microprocessor Power Model

CPU power P_{cpu} consists of:

- dynamic power P_{dyn} ,
- leakage current P_{leak} ,
- short-circuit current P_{sc} ;

Execution Time Model

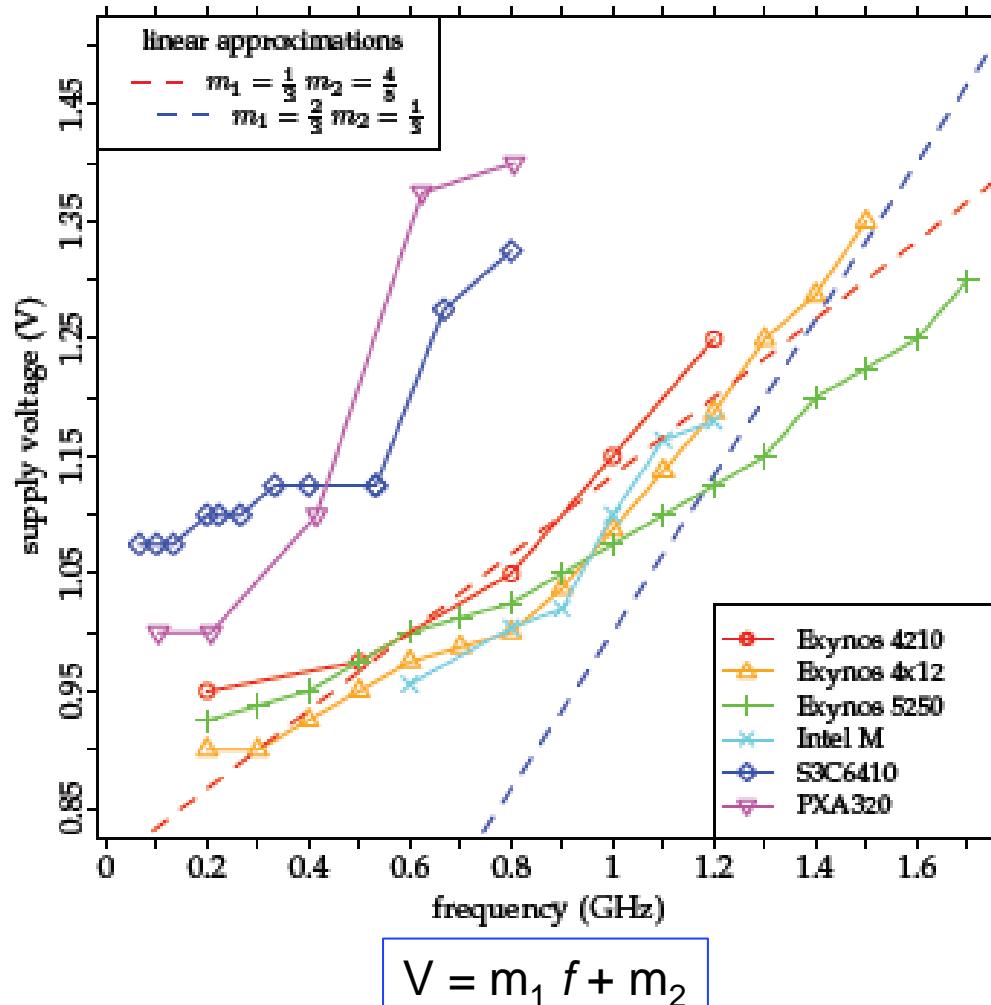
Execution time Δt depends on:

- cc_b code size in clock cycles,
- f CPU clock frequency,
- f_k frequency thieves,
- β slack time per clock cycle;

$$\begin{aligned} P_{\text{cpu}} &= P_{\text{dyn}} + P_{\text{leak}} + P_{\text{sc}} \\ &= (1 + \gamma V) \cdot \eta \alpha C V^2 f \\ &= (1 + \gamma V) \cdot \xi V^2 f. \end{aligned}$$

$$\Delta t = cc_b \left(\frac{1}{f - f_k} + \beta \right).$$

V can be approached by a linear function of the frequency





Optimal frequency and Convexity

- System's energy consumption model (EFCR)

$$\begin{aligned} E_{\text{sys}}(f) &= (P_{\text{cpu}} + P_{\text{back}}) \cdot \Delta t \\ &= ([1 + \gamma V] \xi V^2 f + P_{\text{back}}) \cdot cc_b \left(\frac{1}{f - f_k} + \beta \right), \end{aligned}$$

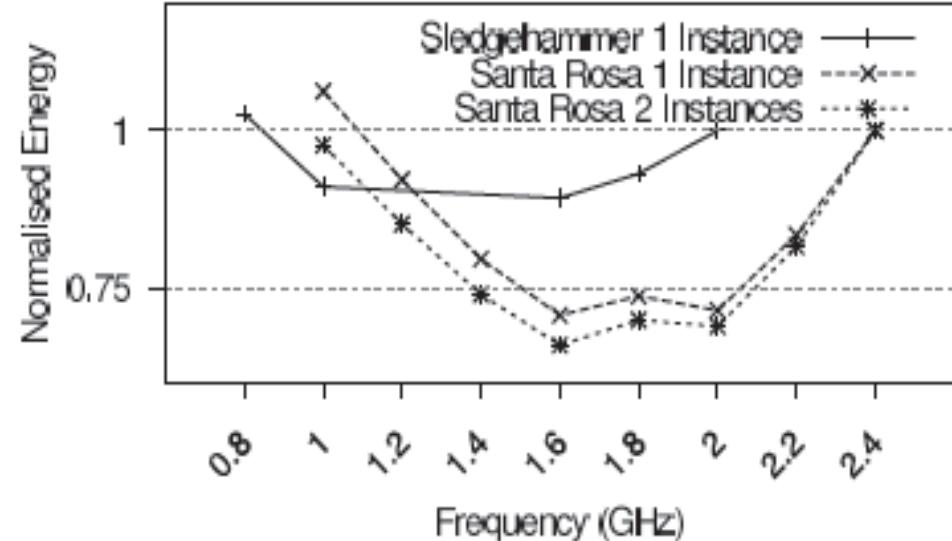
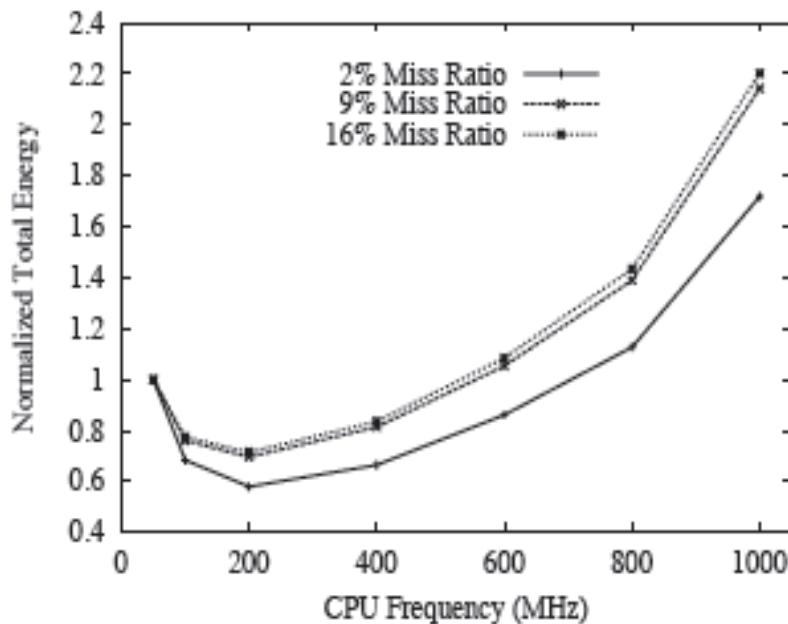
where $\{\gamma, \xi, P_{\text{back}}, cc_b, f_k, \beta\} \in \mathbb{R}^+$.

- A single minimum for $E_{\text{sys}}(f)$ exists at f_{opt} when

$$\left(\frac{\partial E_{\text{sys}}}{\partial f} \right)_{f=f_{\text{opt}}} = 0, \quad \text{and} \quad \frac{\partial^2 E_{\text{sys}}}{\partial f^2} \geq 0 \quad \text{holds;}$$

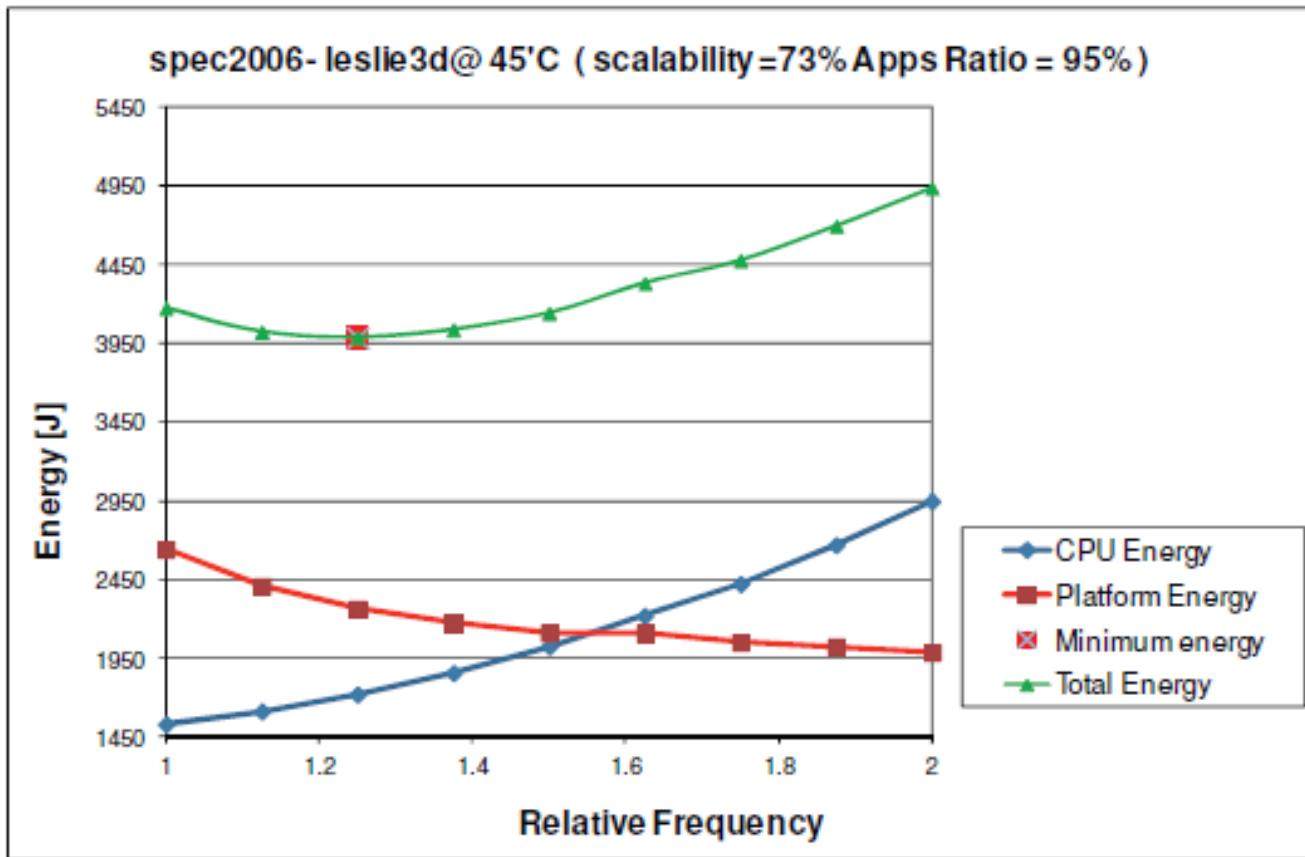
State of the art

- Convexity was already observable, however no analytical studies were performed



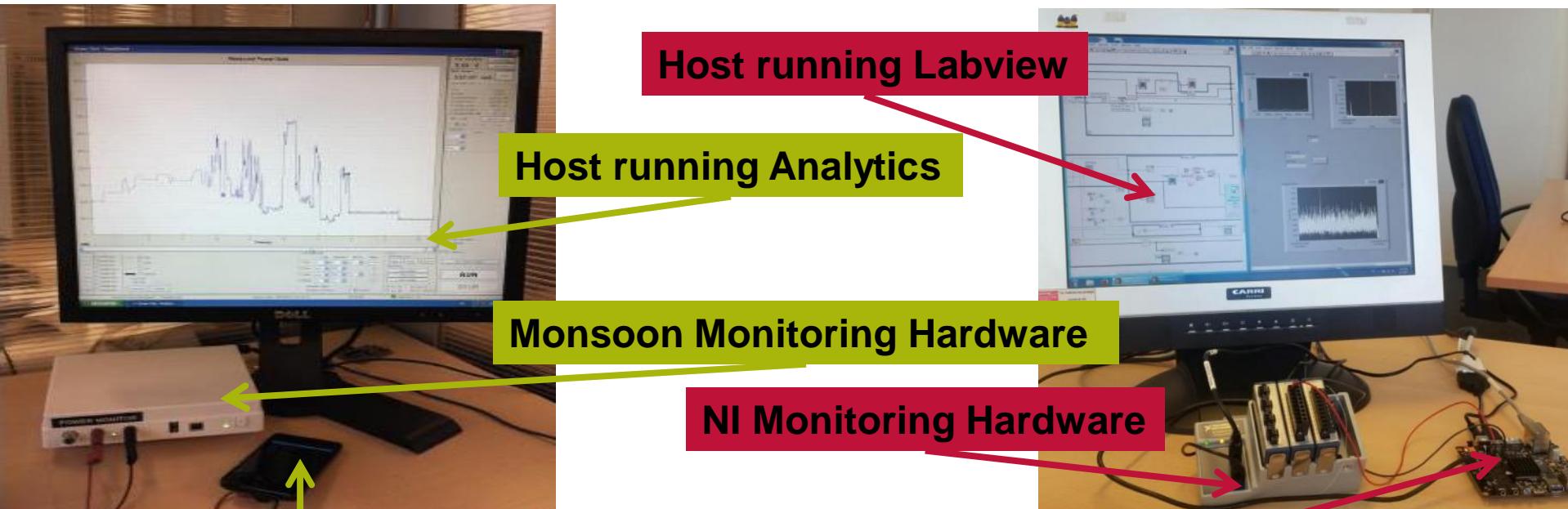
Fan, X., Ellis, C. S., and Lebeck, A. R.
The synergy between power-aware memory systems and processor voltage scaling. In PACS'04
Le Sueur, E., and Heiser, G.
Dynamic voltage and frequency scaling: the laws of diminishing returns. In PACS'10

Convexity shown on Intel Core 2 board



R. Efraim, R. Ginosar, C. Weiser, &A. Mendelson: "Energy Aware Race to Halt A down to EARtH Approach for Platform Energy Management", IEEE Computer Architecture Letter, 2012.

Two Testbeds



Samsung Galaxy SII under test

- Cheaper
- Easier to set up, easier to use
- Probes at the battery level
- Home made analytics

TI AM572x Board under test

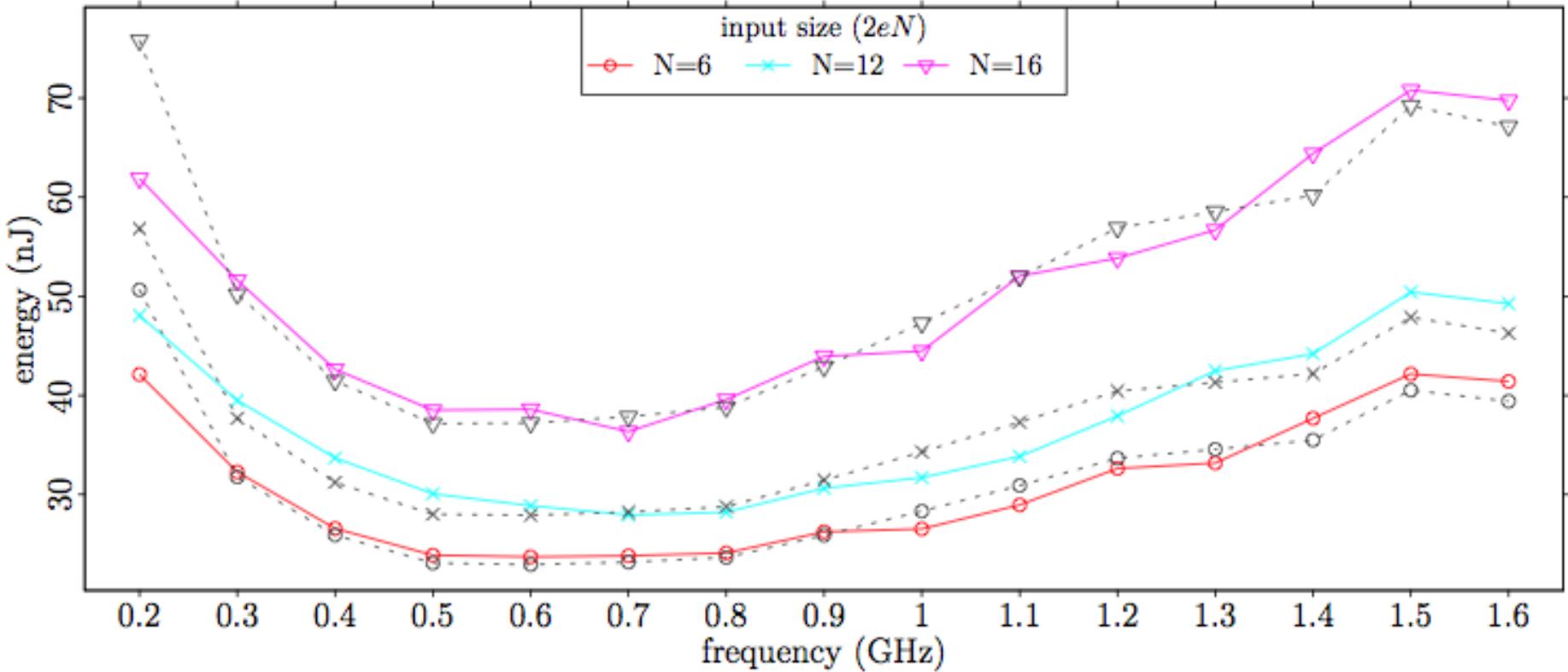
- More expensive
- More accurate
- Probes location anywhere on the board
- Various SW analytics

“Experimental Energy Profiling of Energy-critical Embedded Applications”

K. Vaddina, F. Brandner, P. Jouvelot, and G. Memmi

IEEE SoftCom'17, Split, Croatia, 2017.

Experimental validation with the Samsung smartphone

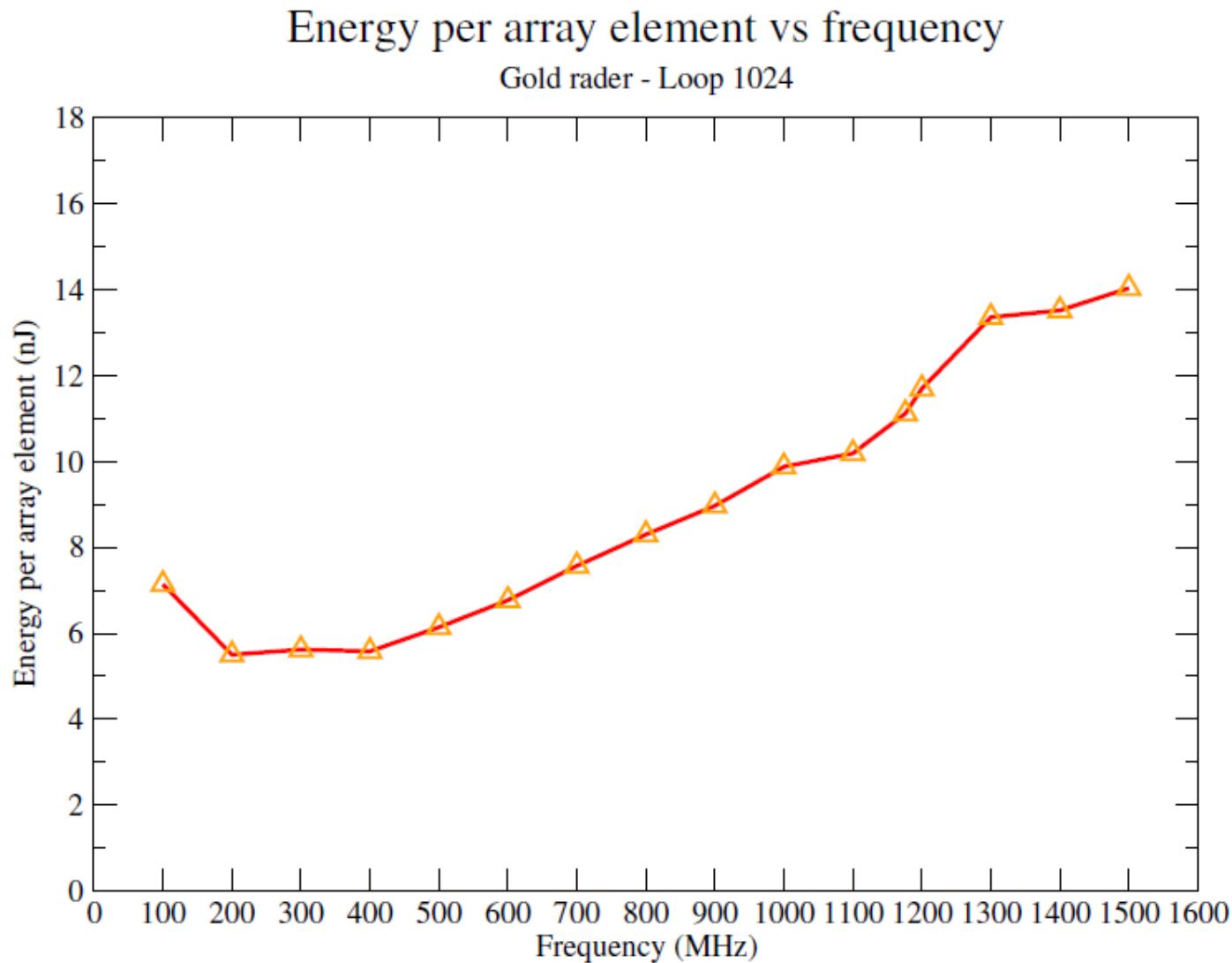


In color: measurements

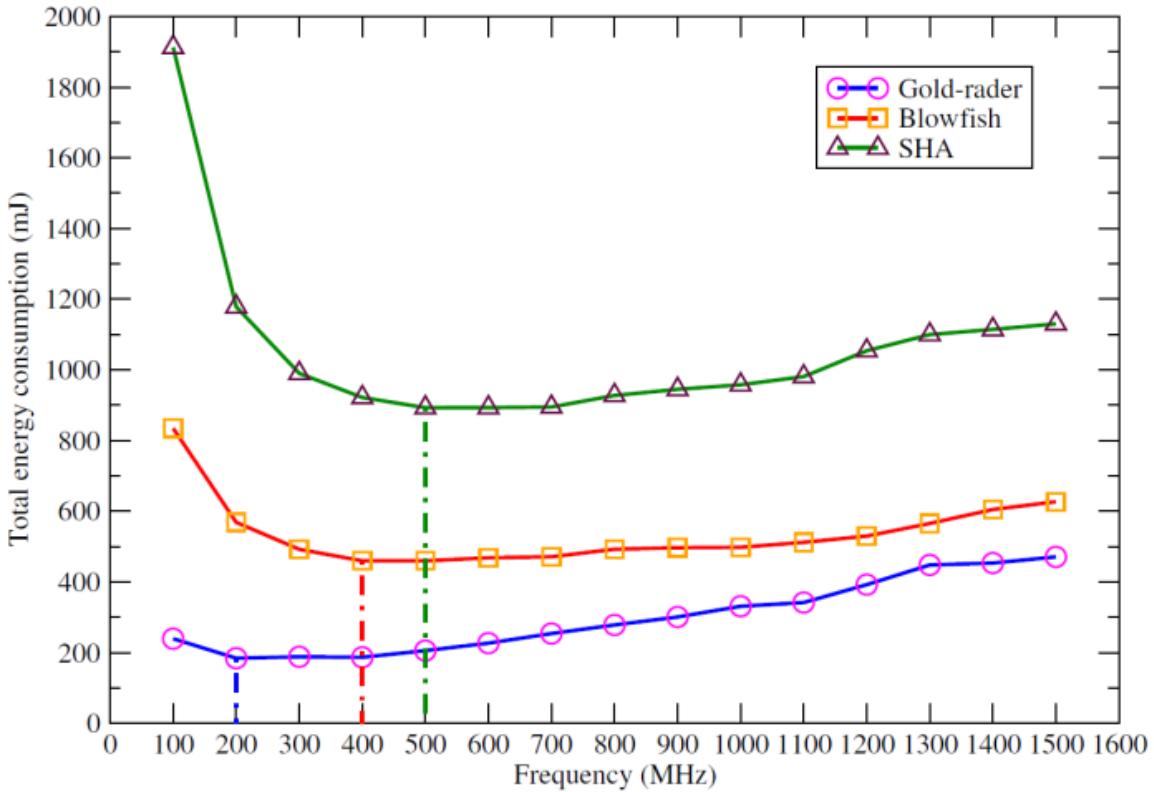
In doted lines: theoretical EFCR calculation

When N increases, f_{opt} stays stable

Experimentation with TI AM572x board



f_{opt} sensitivity



- Energy consumption of three different programs running on TI AM572x platform showing different profiles with different f_{opt} .

“Parameter Sensitivity Analysis of the Energy/Frequency Convexity Rule for Application Processors”

K. De Vogelee, G. Memmi, and P. Jouvelot

J. of Sustainable Computing, Informatics and Systems, Elsevier B.V., September 2017.

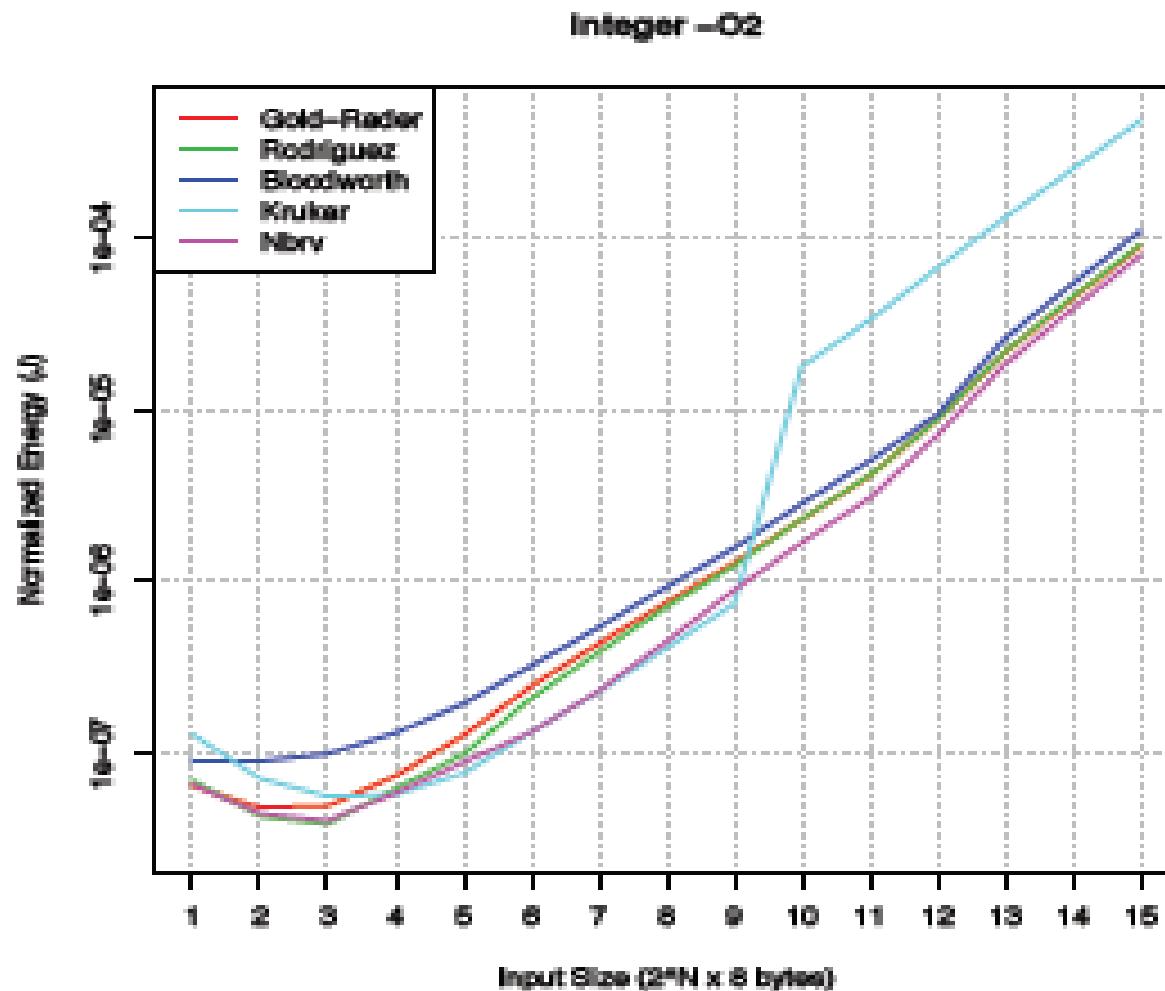


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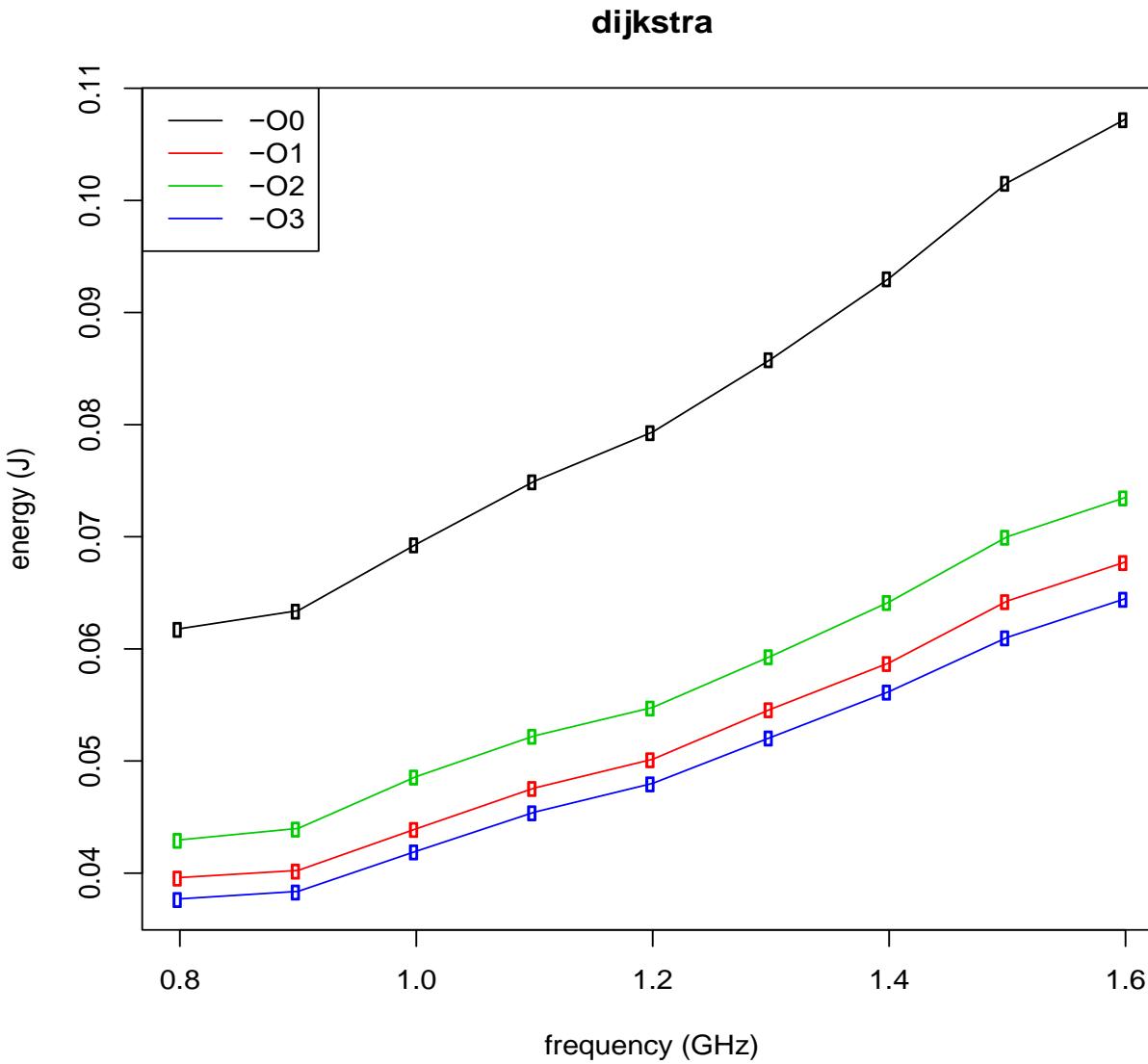
Towards program energy profiling



Energy profiles also can detect anomalies

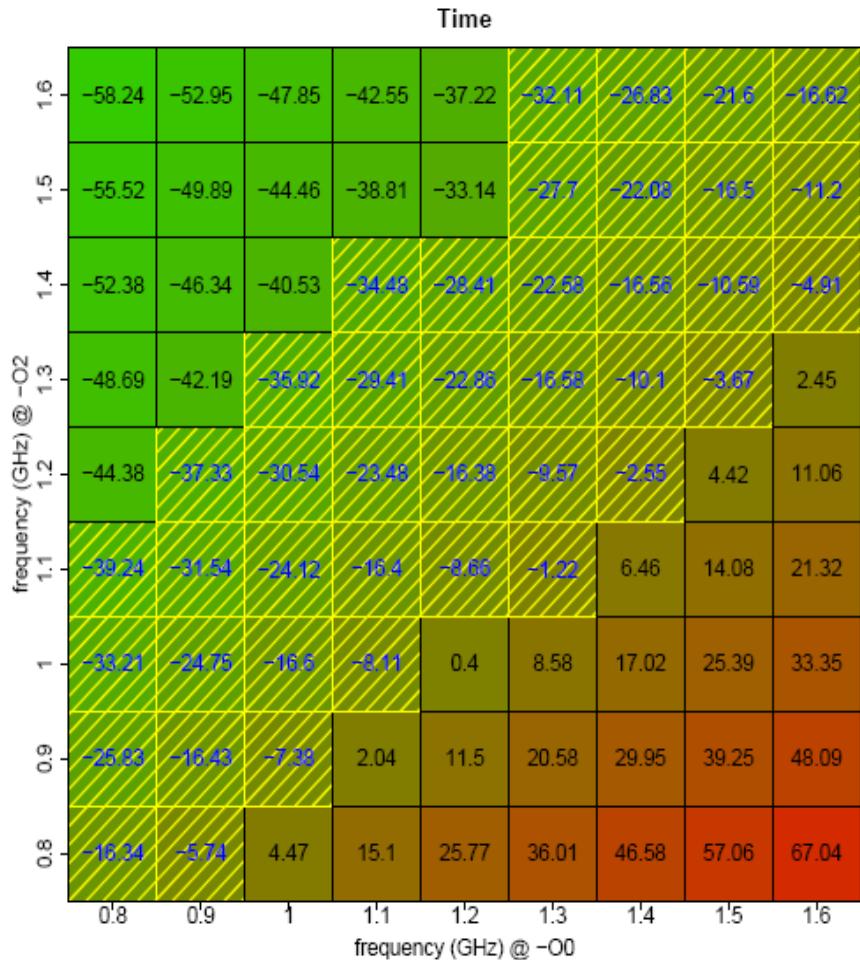
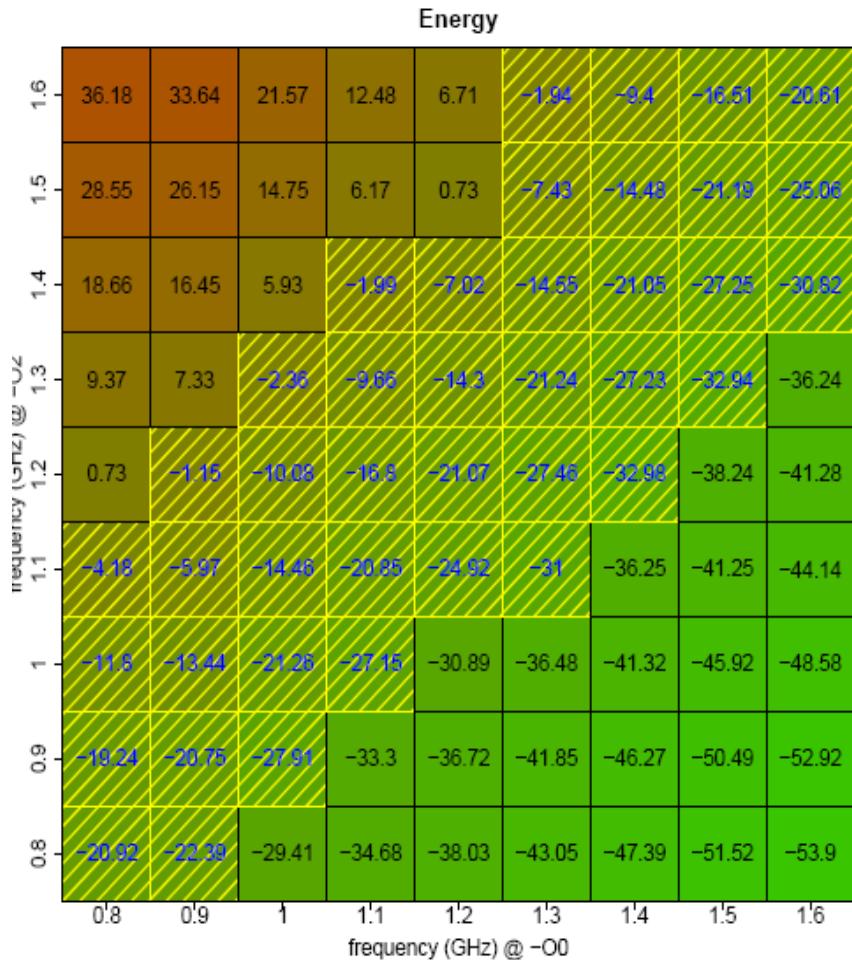


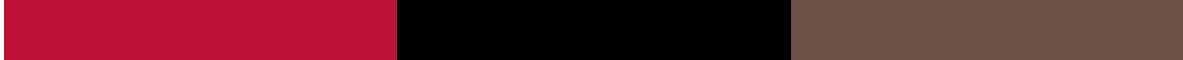
Optimizing for performance also optimizes for energy cpu



Drawing the best from the optimizer and DVFS

- Created by tuning clk frequency and performing standard program transformation





Conclusion

Energy-Oriented Environment

First measurements and results are setting expectations in the 10-40% saving range by:

- ✓ Exploiting energy-frequency convexity
- ✓ Integrating temperature impact in our models

■ Wider array of experimentation

- Using a wider and better controlled temperature range
- Setting a richer and more complete benchmark

■ More research on energy program profiling

- Handling various architectures (e.g. cache)
- Understanding how where, and when to play with clock frequency changes (including overhaed data)
- Temperature online monitoring



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Thank you



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- “Experimental Energy Profiling of Energy-critical Embedded Applications” K. Vaddina, F. Brandner, P. Jouvelot, and G. Memmi *IEEE SoftCom’17, Split, Croatia, 2017.*
- “Parameter Sensitivity Analysis of the Energy/Frequency Convexity Rule for Application Processors” K. De Vogeleeer, G. Memmi, and P. Jouvelot, *J. of Sustainable Computing, Informatics and Systems, Elsevier, September 2017.*
- “Modélisation de la consommation énergétique des programmes : aspects thermiques et loi de convexité énergie-fréquence” K. De Vogeleeer, P. Jouvelot, and G. Memmi, *ICSSEA’16 then Génie Logiciel 117 pp 47-59, June 2016.*
- “Modeling Temperature Bias of the Power Consumption of Nanometer-Scale CPUs in Application Processors.” K. De Vogeleeer, G. Memmi, P. Jouvelot, and F. Coelho *International Conference on Embedded Computer Systems: Architectures, Modeling, and Simulation, SAMOS XIV, July 2014.*
- “The Energy/Frequency Convexity Rule: Modeling and Experimental Validation on Mobile Devices” K. De Vogeleeer, G. Memmi, P. Jouvelot, and F. Coelho *10th International Conference on Parallel Processing and Applied Mathematics, PPAM 2013, PEAC Workshop on "Power and Energy Aspects of Computation", Warsaw, Poland, pp 793-803, September 2013.*