

TELECOM  
ParisTech



MINES  
ParisTech

# Thermal behavior and Energy/Frequency Convexity Rule of Energy Consumption for Programs

Karel De Vogeleer  
Pierre Jouvelot  
G rard Memmi



# Motivations for focussing on mobile computing

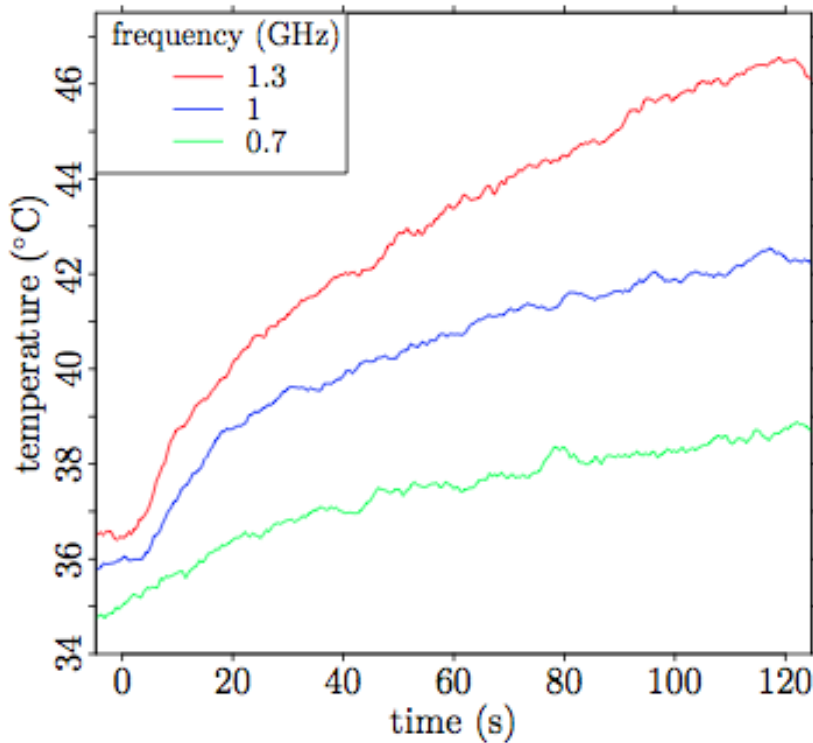
- It is **NOT** the energy saving per se :
  - A smartphone CPU consumes between 60 to 400mW
  - There are about  $7 \times 10^9$  smartphones sold in the last 5 years, there will be  $50 \times 10^9$  'smart objects' in 2020
  - A saving of 30% would provide grossly about 280 MW for the smartphones, about 3 GW for the smart objects
  - This would only save between a tidal and a nuclear power station
- Focussing on mobile systems: they are '*energy-critical*' : it is being constantly looking for providing more autonomy with a QoS unchanged
- It is about a natural-resource-free energy saving



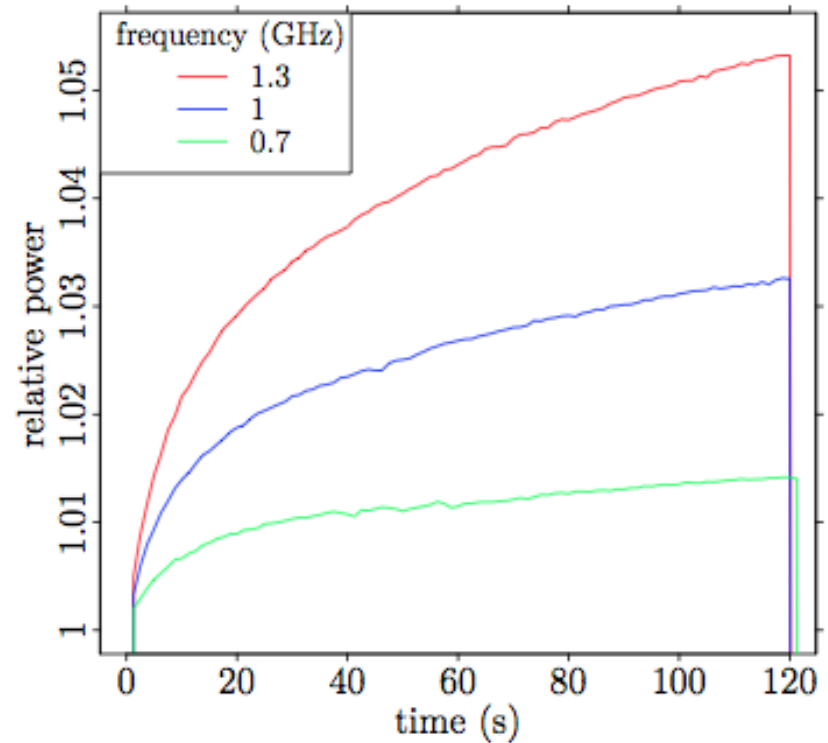
# Thermal Behavior: Power-temperature rule Passive Cooling rule

# Temperature impacts energy consumption

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



(a) température



(b) puissance

# Passive Cooling Rule

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

internal heat generation + radiation + convection

$$mC \frac{dT}{dt} = P(T) + \epsilon\sigma(T^4 - T_a^4) + h(T - T_b),$$

which has the solution

$$t(T) = -\frac{1}{\kappa_4} \left( A \ln |T - \omega_1| + \frac{C}{2} \ln |(T - \alpha)^2 + \beta^2| \right. \\ \left. + B \ln |T - \omega_2| + \frac{\alpha C - D}{\beta} \arctan \left( \frac{T - \alpha}{\beta} \right) + c_0 \right).$$

# Approximative solutions

- Coefficient approximation:

$$T = \frac{\omega_1 \pm \omega_2 c_0 e^{-\frac{\kappa_2}{A} t}}{1 \pm c_0 e^{-\frac{\kappa_2}{A} t}};$$

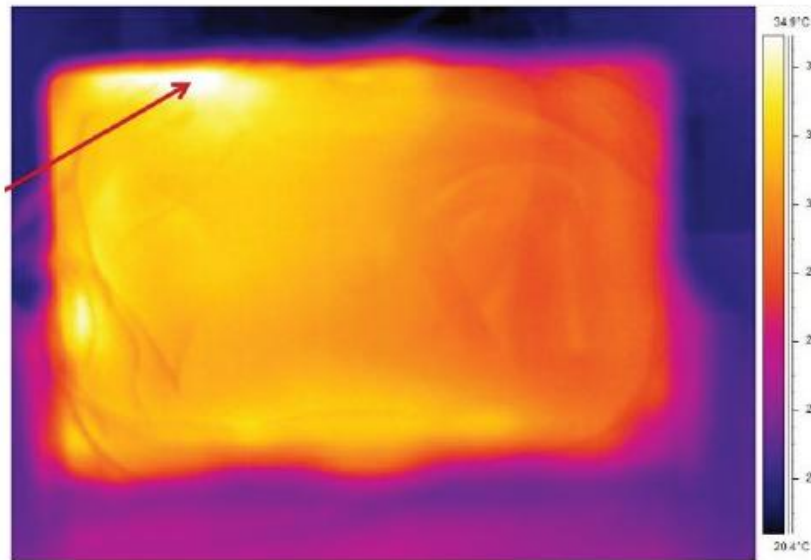
- First-order O'Sullivan:

$$T = \left( T_0 - T_a + \frac{p}{n} \right) e^{-\frac{n}{c} t} - \frac{p}{n} + T_a;$$

- Second-order O'Sullivan:

$$T = \frac{w}{2m} (\text{co}) \tanh \left( \frac{w}{2m} t + c_0 \right) - \frac{n}{2m} + T_a.$$

# Isothermal assumption



(a) an Apple iPad by Wagner and Maltz [20]

(b) an unnamed thin notebook by Wagner and Maltz [21]

Fig. 1. Experimental thermal imaging of the skin temperature of (a) an Apple iPad [20] and (b) an unnamed thin notebook [21]. The iPad exemplifies the quasi isothermal surface of an embedded system. The surface temperature varies between 30°C and 35°C. On the other hand, the thin notebook shows large temperature variations, between 48°C and 25°C. Here, aggressive active cooling methods extract the heat as fast as possible from the heat sources inside the device.

# Contribution on thermal behavior

- **Necessary for reproducible measurement and for accurate energy consumption models**
- **Power – temperature relationship**
- **Approximations for practical use**





# EFCR: the energy – frequency convexity rule

# Fragmenting energy consumption per system module

- System's energy consumption  $E_{\text{sys}}$  definition

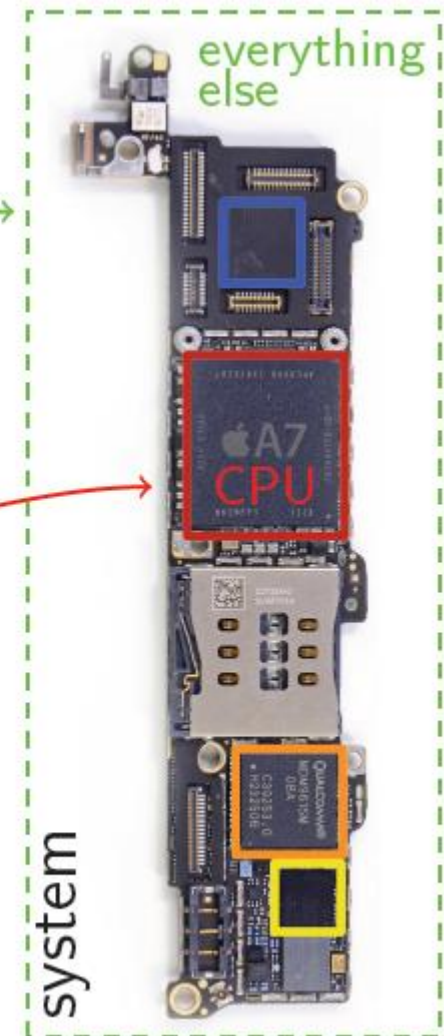
$$\begin{aligned} E_{\text{sys}} &= \int_0^{\Delta t} P_{\text{total}} dt \\ &= \int_0^{\Delta t} (P_{\text{cpu}} + P_{\text{back}}) dt; \end{aligned}$$

- Examples of  $P_{\text{back}}$  include:

- ▶ LCD screen,
- ▶ radio interface,
- ▶ power supply;

- If  $P_{\text{cpu}}$  and  $P_{\text{back}}$  are constant over  $\Delta t$ :

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



# Power and time model

## Microprocessor Power Model

CPU power  $P_{\text{cpu}}$  consists of:

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

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

## Execution Time Model

Execution time  $\Delta t$  depends on:

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

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

# Optimal frequency and Convexity

- System's energy consumption model (EFCR)

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

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

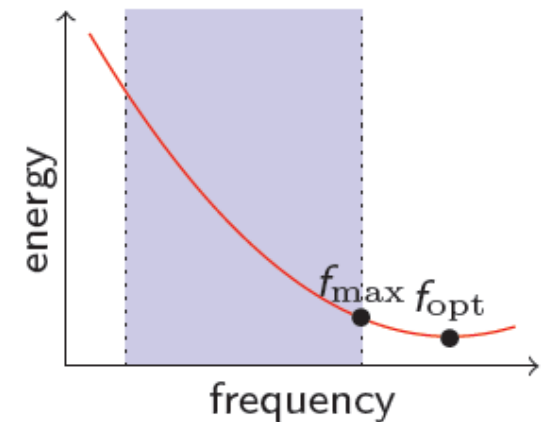
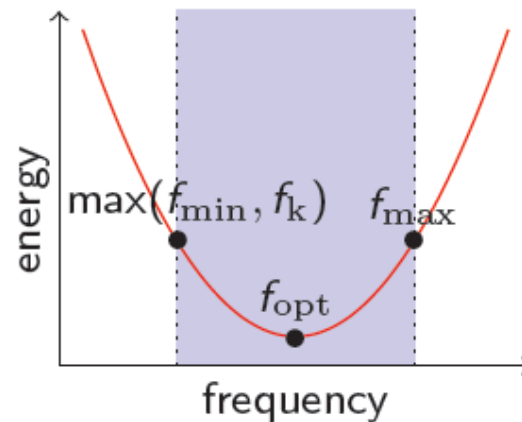
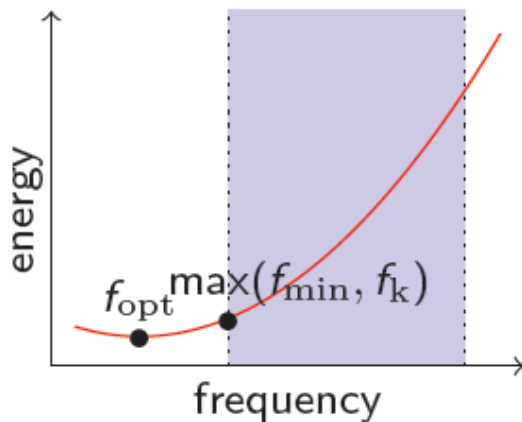
- A single minimum for  $E_{\text{sys}}(f)$  exists at  $f_{\text{opt}}$  when

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

- $V$  is approximately an affine map of  $f$ :  $V \rightarrow m_2 f + m_1$ .

# 3 classes of processors

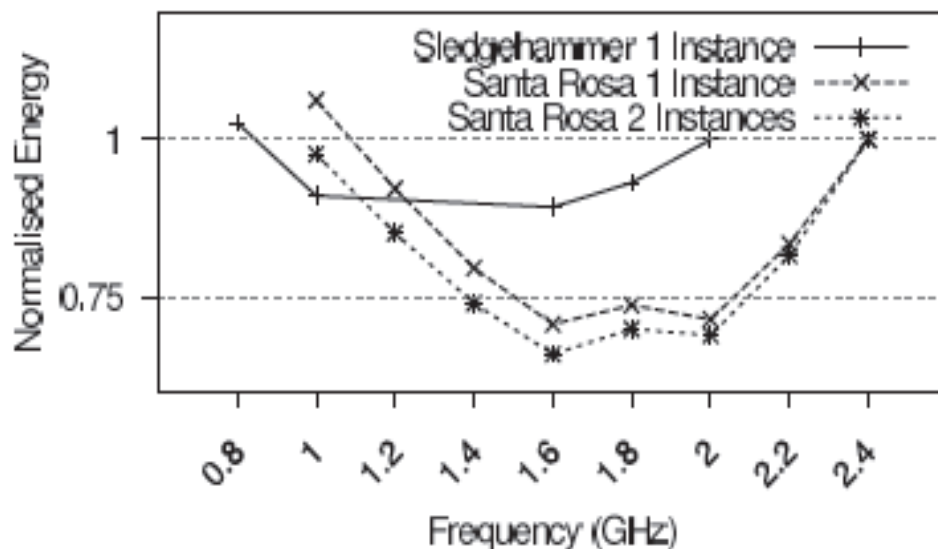
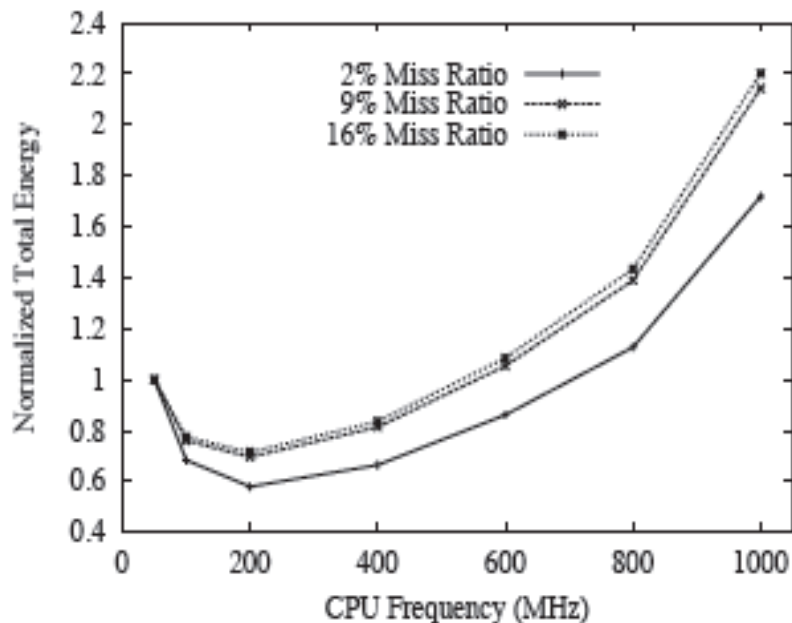
- **There exists a domain of frequencies where the processor is delivered by the manufacturer:  $f_k$  must be smaller than  $f$  since  $f_k$  a fraction of  $f$**
- **Save for overclocking or underclocking**



1	$f_{opt} < \max(f_{min}, f_k)$	the slower the better
2	$\max(f_{min}, f_k) \leq f_{opt} \leq f_{max}$	chase $f_{opt}$
3	$f_{max} < f_{opt}$	race-to-halt

# State of the art

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



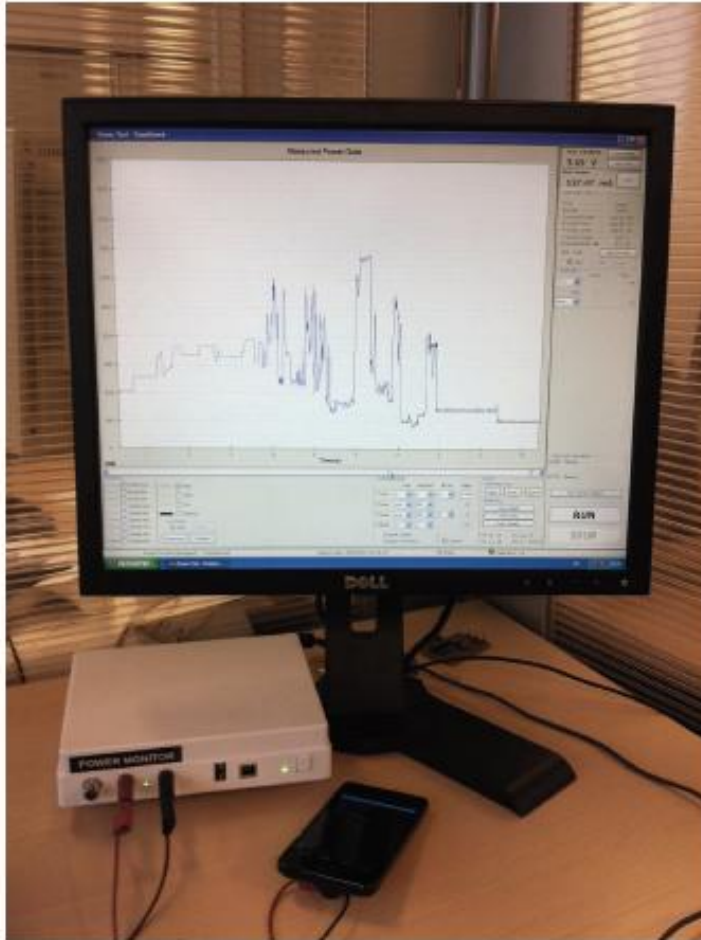
Fan, X., Ellis, C. S., and Lebeck, A. R.

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

# Testbed

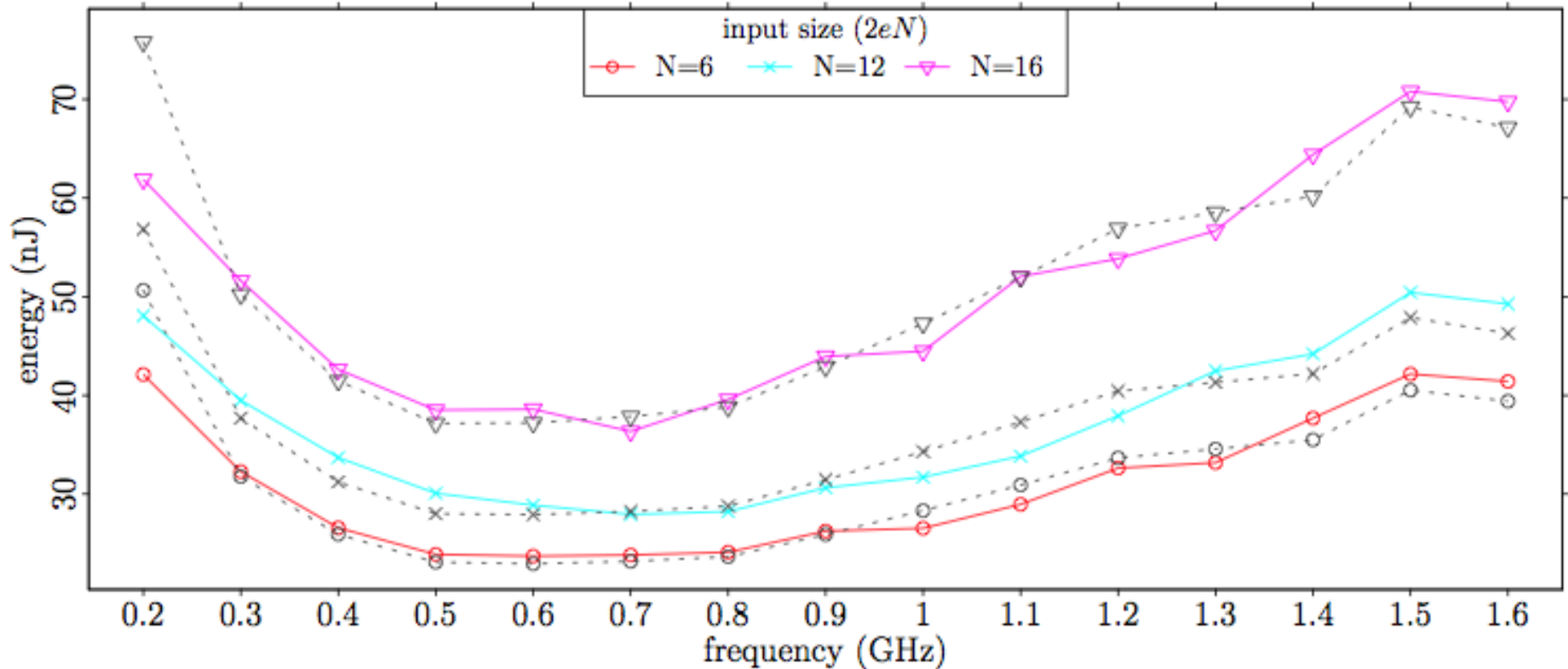


- **Benchmark:** bit-reverse algorithm, part of the FFT algorithm:

```
void bitreverse_gold_rader
  (int N, complex *data) {
  int n = N, nm1 = n-1;
  int i = 0, j = 0;
  for (; i < nm1; i++) {
    int k = n >> 1;
    if (i < j) {
      complex temp = data[i];
      data[i] = data[j];
      data[j] = temp;}
    while (k <= j) {
      j -= k; k >>= 1;}
    j += k ;
  }
}
```

- **Testbed:** Samsung Galaxy SII;
- **Power Measurement:** Monsoon.

# Experimental validation



In color, the measurements

In dotted lines the theoretical EFCR calculation

When  $N$  increases,  $f_{opt}$  stays stable

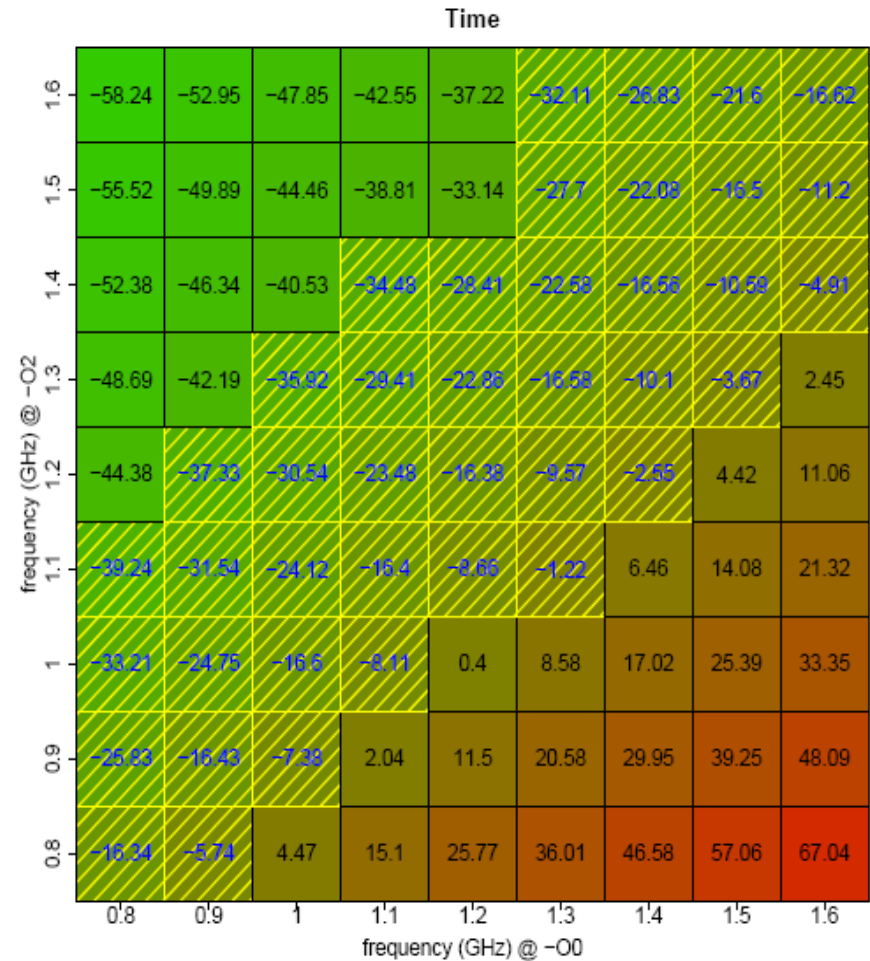
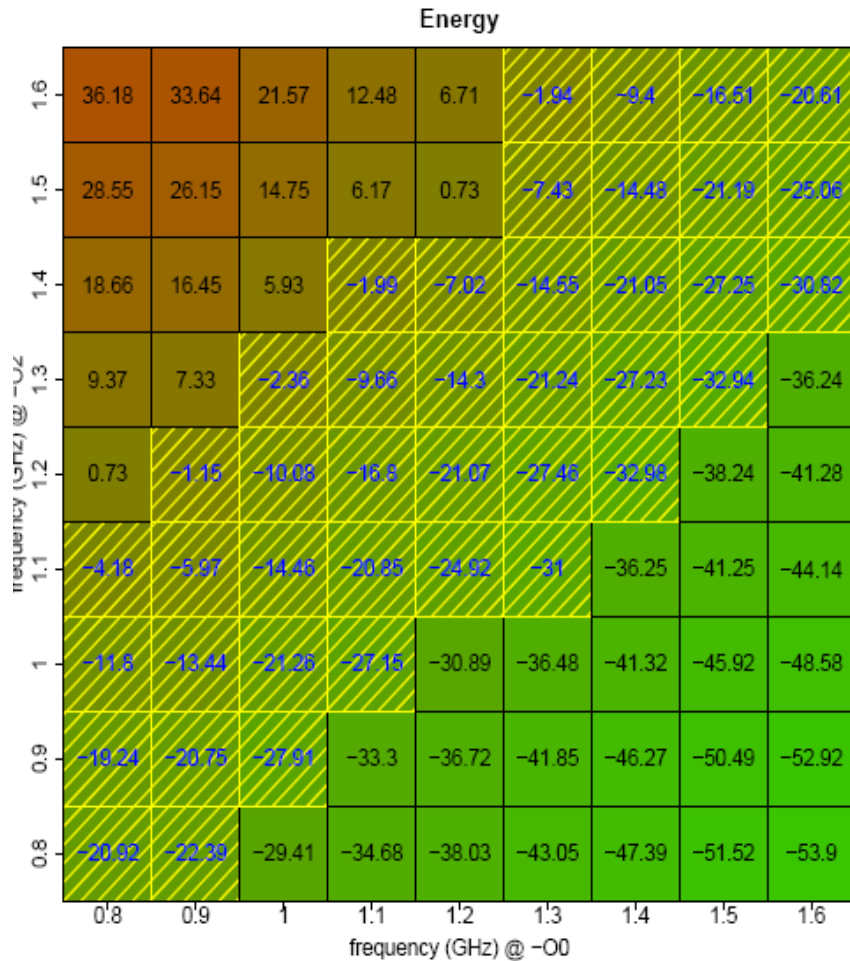




# Conclusion

# Towards energy program profiling

- Created by tuning clk frequency and performing standard program transformation



# Energy-Oriented Environment

First measurement and results are very encouraging, setting expectations in the 10-40% saving range rather than in the 2-5% initially anticipated saving range

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

- **Wider array of experimentation, more accurate measurement and mathematical models, wider temperature range**
- **More research on energy program profiling**
  - Handling various architectures (eg cache)
  - Understanding how and when to play with clock frequency changes
  - Temperature on line monitoring

**Towards an energy-oriented compiler middle-end, and operating system technology**



# Bibliography

- K. De Vogeleer, G. Memmi, P. Jouvelot, and F. Coelho, “The Energy/Frequency Convexity Rule: modeling and experimental validation on mobile devices,” in *Proceedings of the 10th Conference on Parallel Processing and Applied Mathematics*. Springer Verlag, Sep. 2013.
- K. De Vogeleer, G. Memmi, P. Jouvelot, and F. Coelho, “Modeling the temperature bias of power consumption for nanometer-scale CPUs in application processors,” in *14th International Conference on Embedded Computer Systems: Architectures, Modeling, and Simulation*, Jul. 2014, pp. 172-180.
- K. De Vogeleer, P. Jouvelot, and G. Memmi, “The impact of surface size on the radiative thermal behavior of embedded systems,” CoRR, vol. abs/1410.0628, 2014, (submitted to *IEEE TMC* in 2014).



Institut  
Mines-Télécom

# Thank you

