



IRIT

Using ANN for power consumption estimation

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Journée Deep Learning @ IRIT

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Improving energy-efficiency of ICT

Knowledge is Power

Power consumption knowledge is crucial

- Datacenter operators
 - Power capping
 - Billing in colocation centers
 - Improving energy efficiency using autonomous systems
- Users and developers
 - Increasing their awareness
 - Choosing the best application or library



Two techniques:

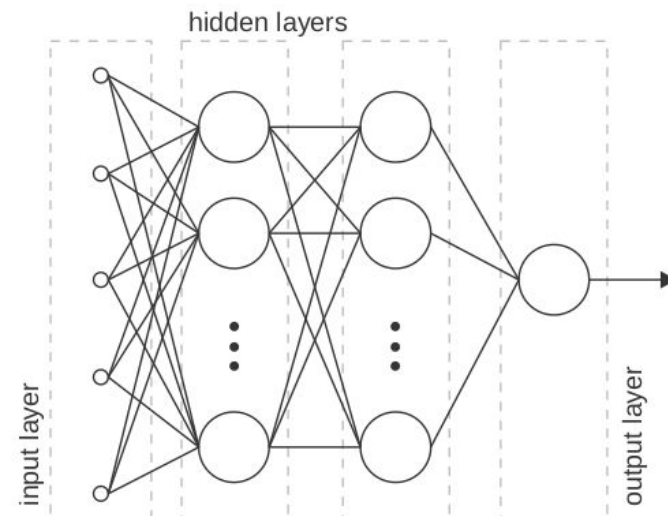
- Power meters
 - Costly
- Power models
 - Analytical models based on DataSheets
 - Regression models
 - Neural Networks



Multi-scale models

Power consumption through the ages

- Linear classical model: error ~ 10-15%
 - $Power = P_{min} + Load \times (P_{max} - P_{min})$
- Adding frequency and voltage: error ~ 5-9%
 - $Power = P_{min} + Load \times \alpha Voltage^2 Frequency$
- Adding temperature: error ~ 4-7%
 - $Power = P_{min} + Load \times \alpha Voltage^2 Frequency + \lambda Temperature$
- Adding PSU bias, cooling, ...: error ~ 2-3%
 - $Power_{DC} = \omega_0 + \omega_1 Power_{AC} + \omega_2 Power_{AC}$
- Learning method
 - Neural network: error ~ 2%

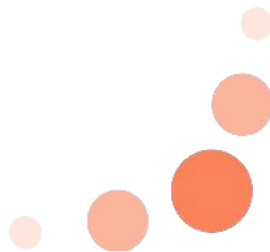


The logo for IRIT, consisting of the letters 'i', 'R', 'I', and 'T' in a white, sans-serif font, set against a large orange circular background. The 'i' has a white dot above it.

IRIT



- 1. Acquiring data**
- 2. Tuning the network**
- 3. Results**



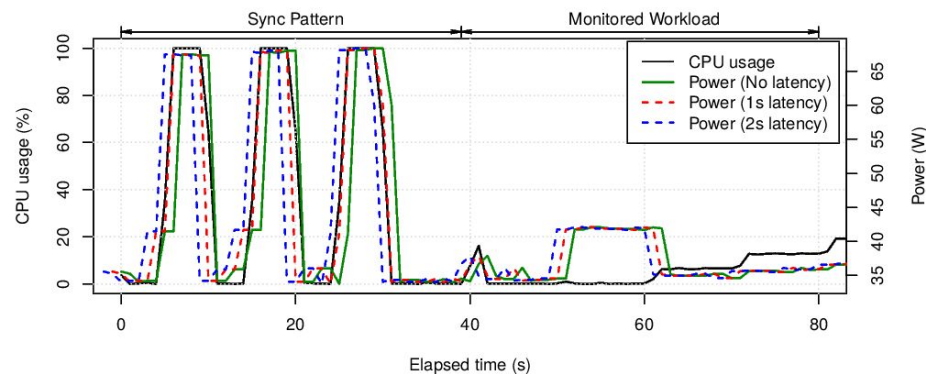
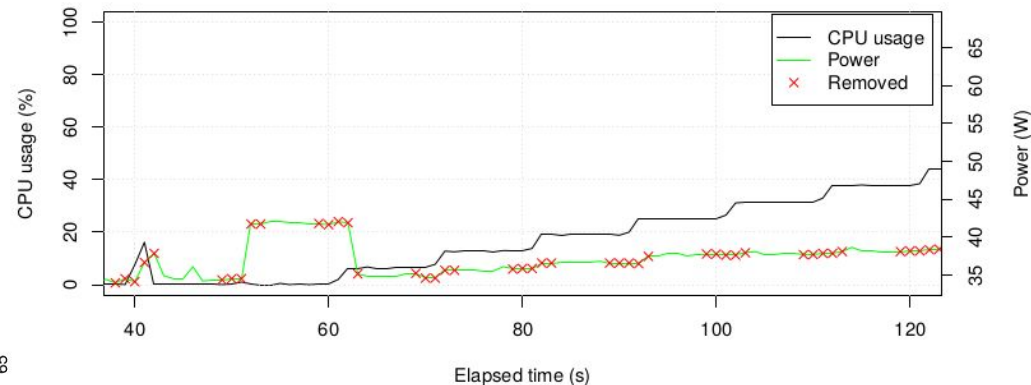
Clapperboard are useful

Direct power measurement is difficult

- Raw power data are noisy
- Time and value precision are low
- High overhead of measures

Several methods to cope with imprecision

- Reaching steady states
- Reduce measure points
- Use movie-like Clap

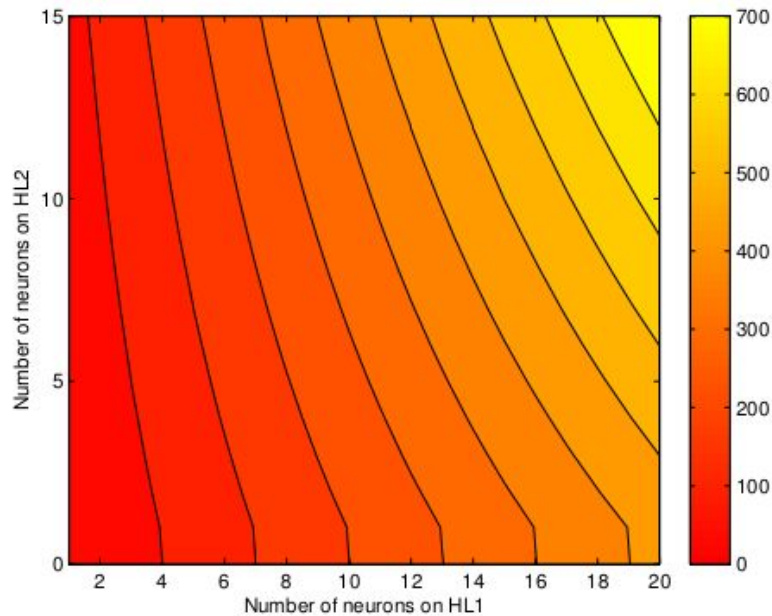




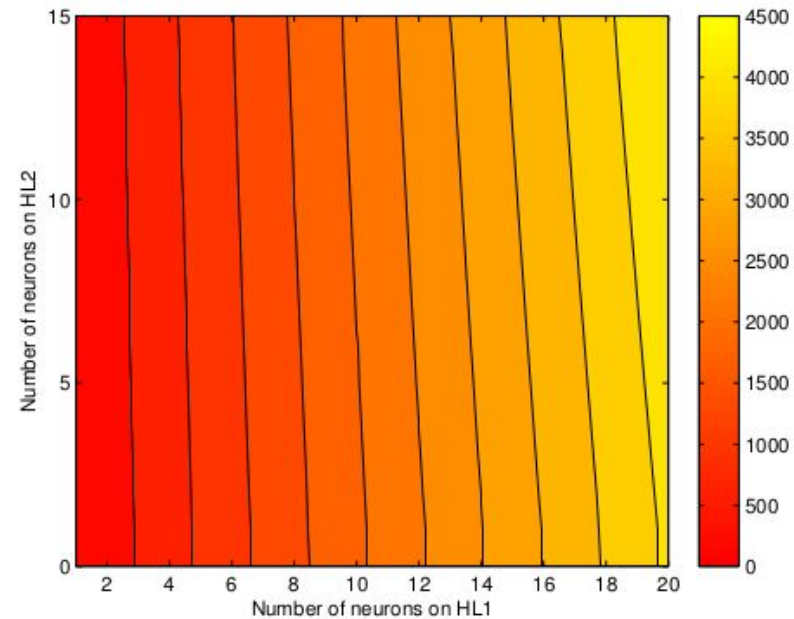
Network topology

Multi-layer perceptron (MLP) with one or two hidden layers

- Number of weights and biases to optimize:
 - $wb = (i + 1) * l_1 + (l_1 + 1) * l_2 + (l_2 + 1) * o$
 - i : inputs; o : outputs
 - l_1 and l_2 neurons on hidden levels



(a) 20 dimensions



(b) 200 dimensions



Variable reduction

Input from more than 250 system-level sensors

One experiment : 1.3 Go/node

Multiple methods for variable reduction

- Increasingly add the most correlated variable
- Increasingly add the variable correlated with residual
- Increasingly add the best (using ANN) variable

Need of a stop criteria

- Adding a variable stops improving the ANN

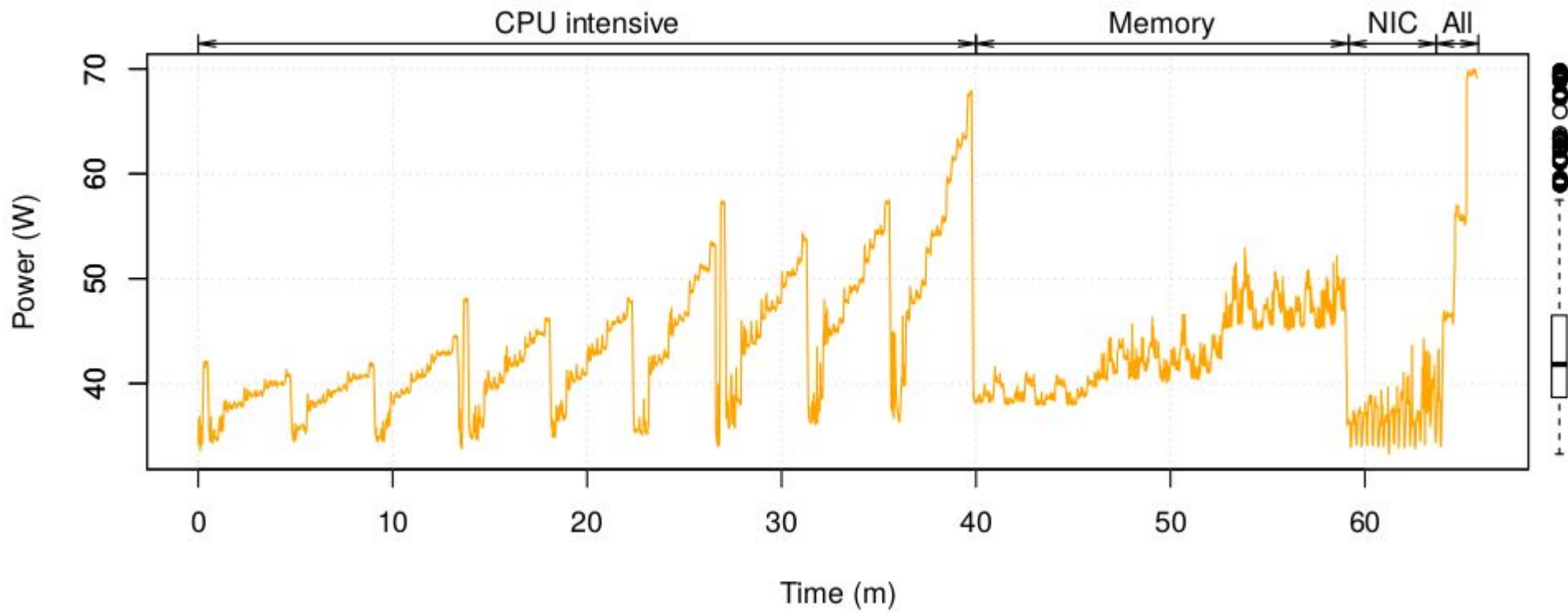




Validation

Coverage problem

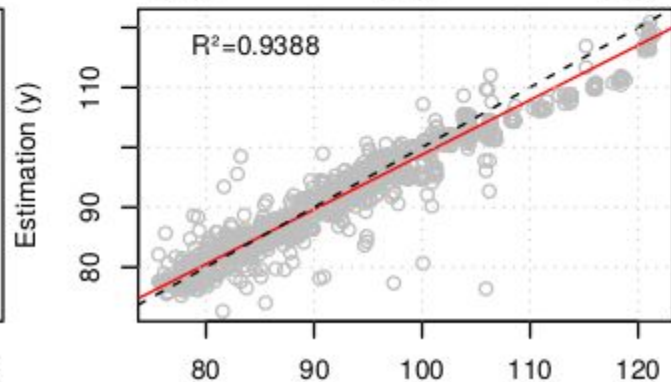
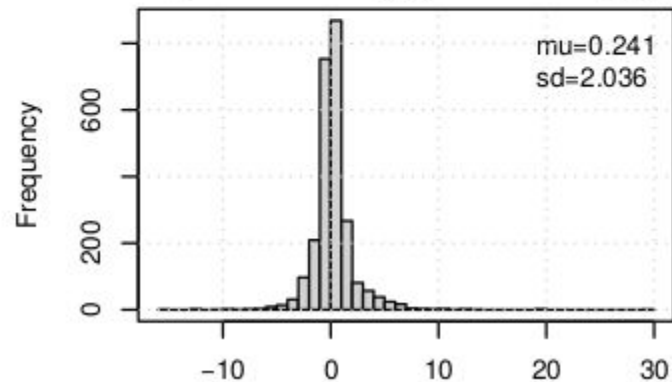
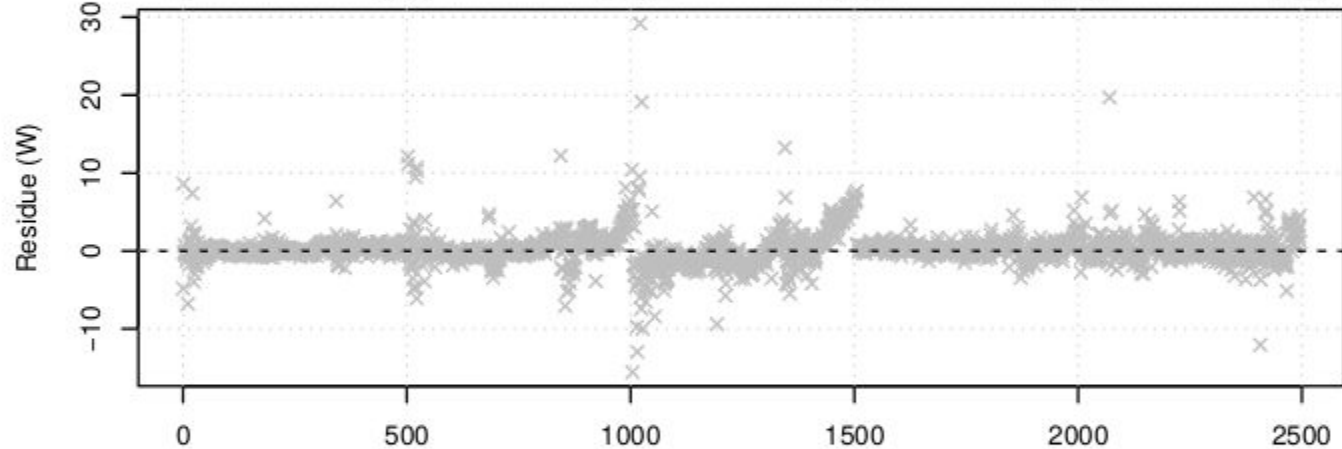
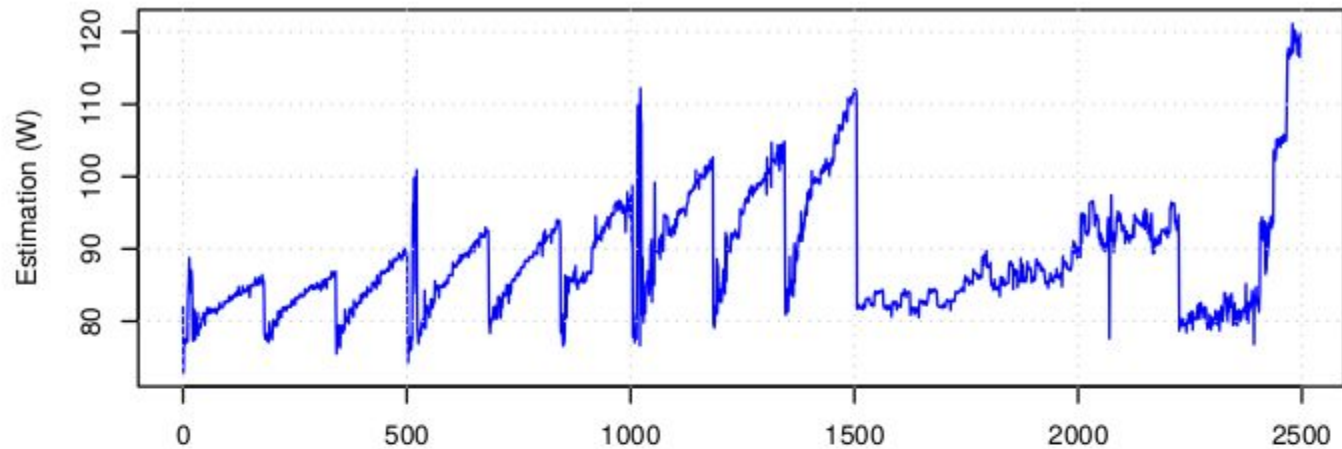
- ANN cannot guess
- Problem of overfitting





Results

- Validation with similar inputs
- With bias removal
- With optimized parameters



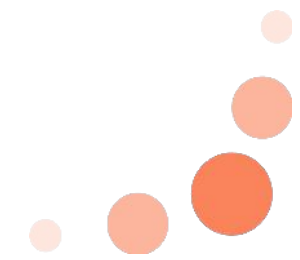
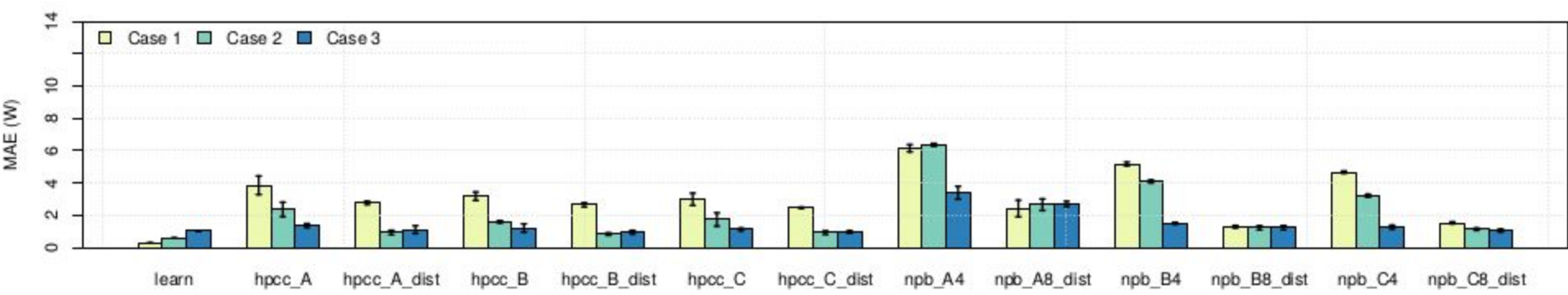
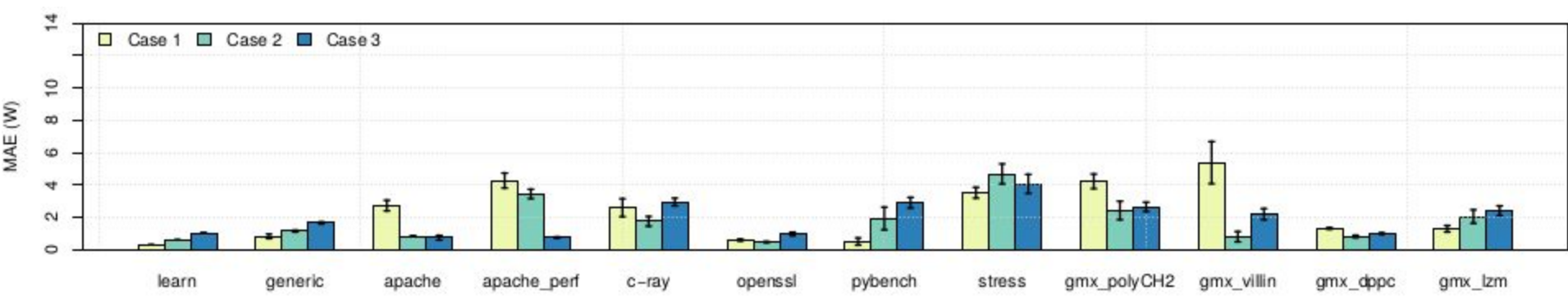
Residue (W)

Target (t)



Performance evaluation

Using workload re-execution





Perspective for SEPIA team

- **Advanced users**
 - Tune models to reduce their impact
 - Use ANN as $\mathbb{R}^n \mapsto \mathbb{R}$
- **Usecases**
 - Power models
 - Prediction models
 - Power
 - Time
 - Crisis management: similarities detection for rapid-flooding
 - Short term: real-time alert
 - Long term: improve area resilience
 - Smart cities and Datacenter optimization

