

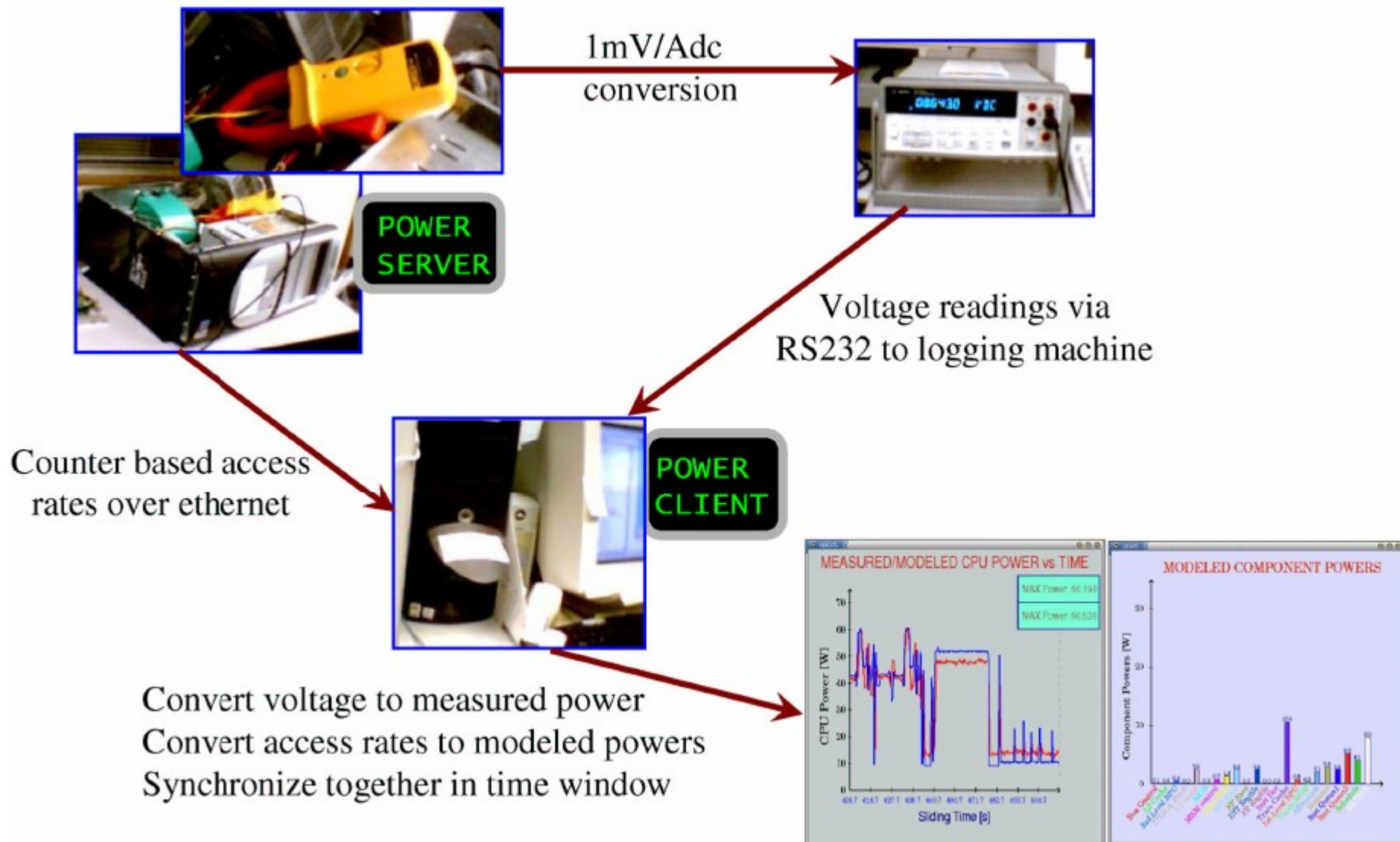
Identifying Program Power Phase Behavior Using Power Vectors

Canturk Isci and Margaret Martonosi
Princeton University

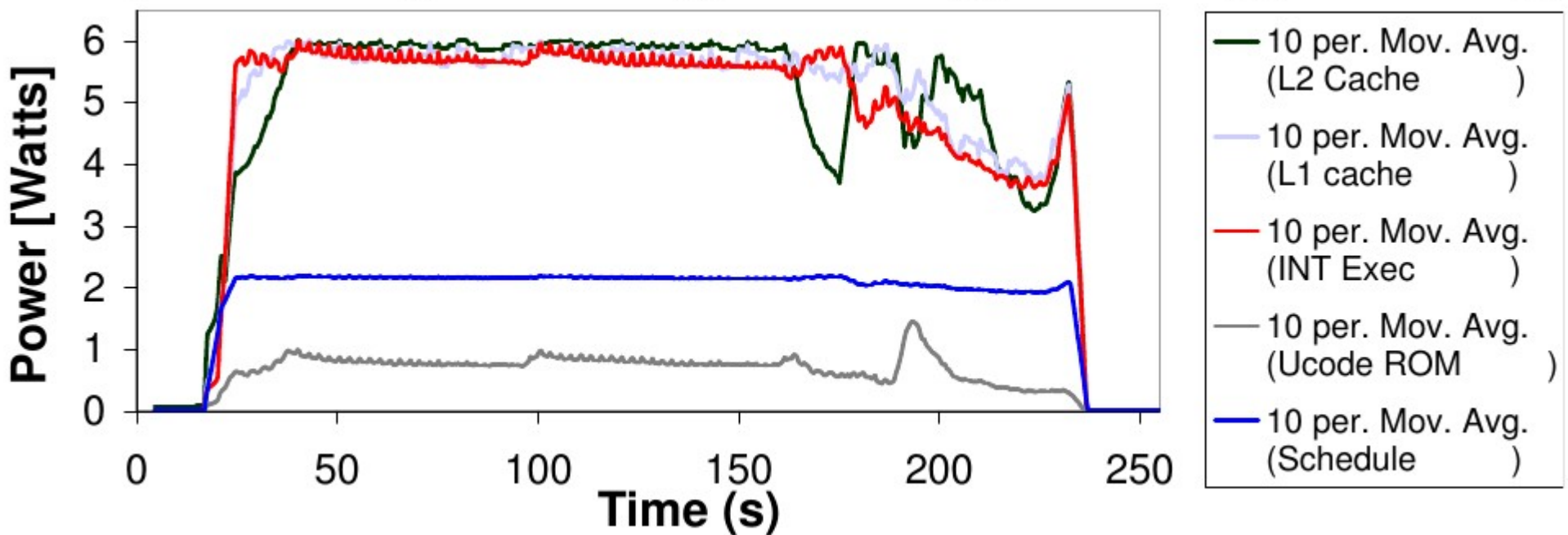
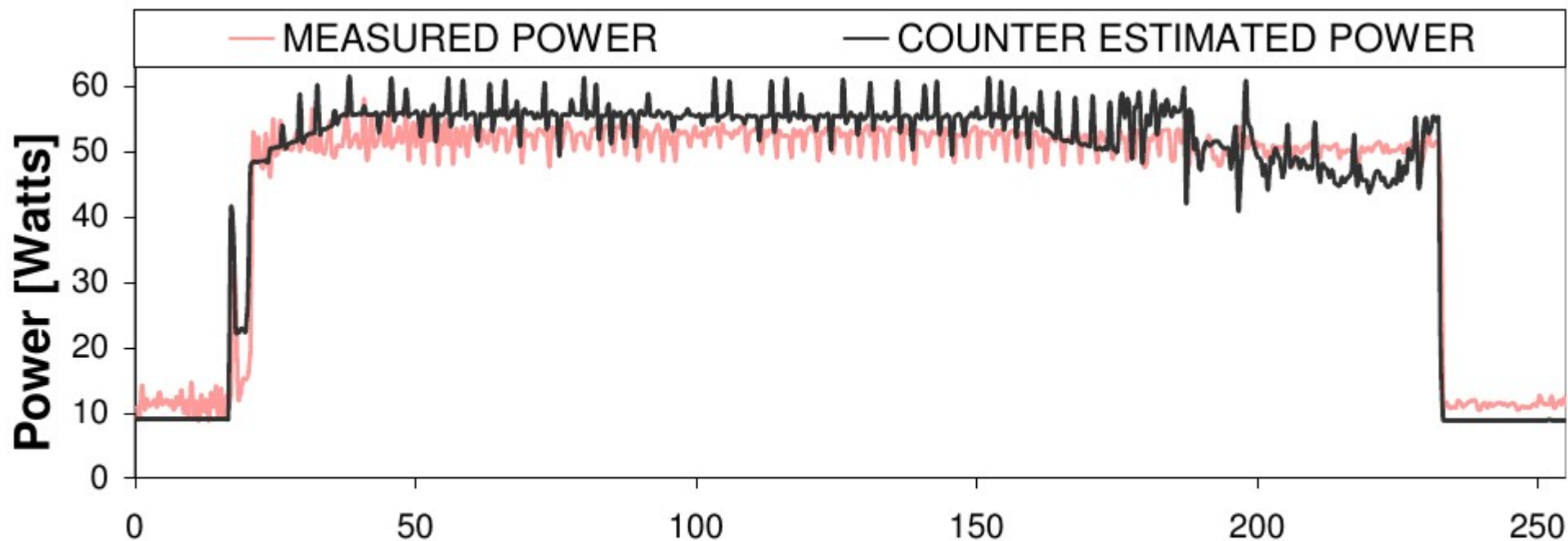
Overview

- A phase analysis method related to power
- Estimates for power values of 22 processor components
- Power vectors acquired at runtime
- Power signatures of programs
- Representative execution points usable for simulation

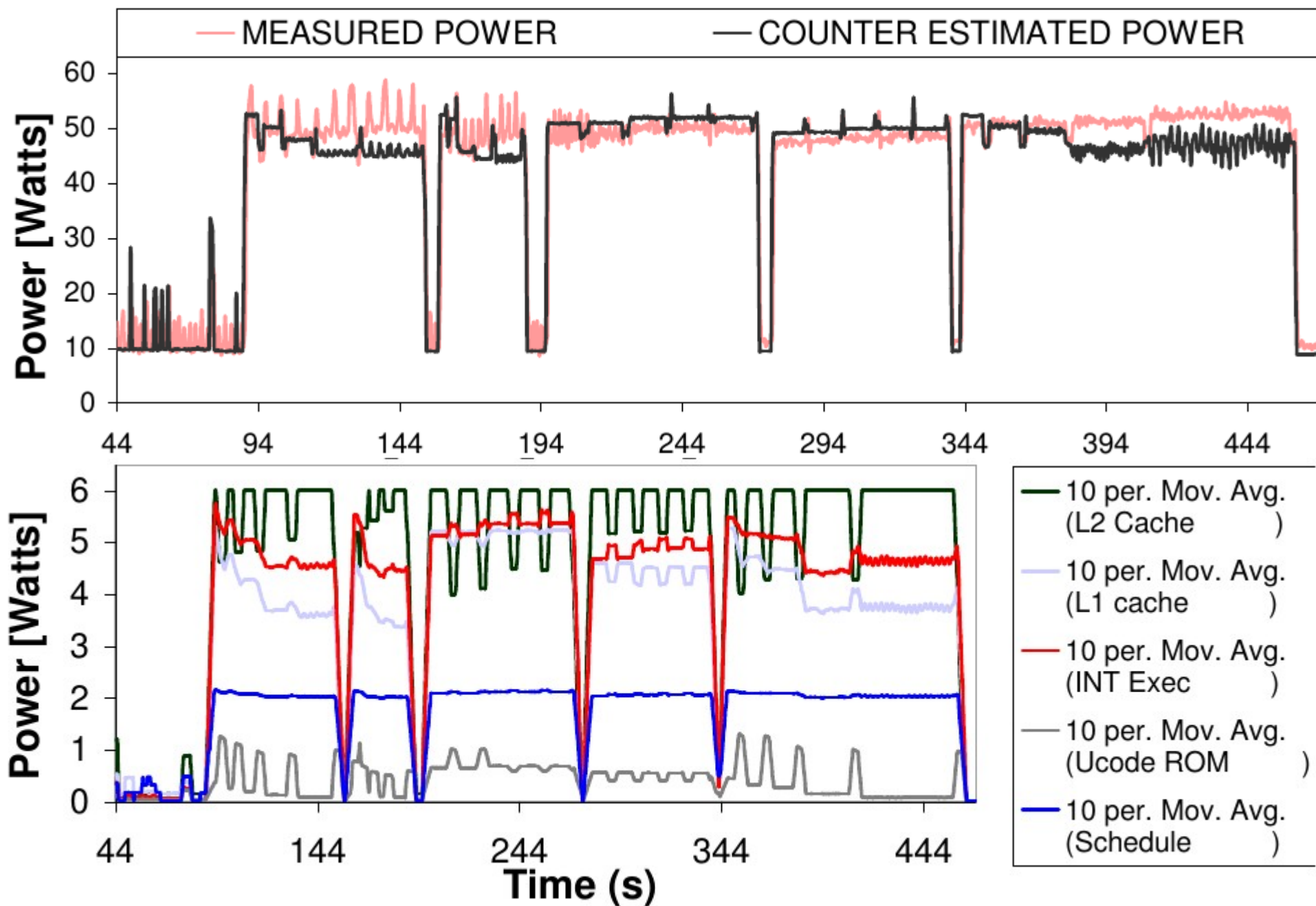
Methodology



Observations (gap)



Observations (gzip)

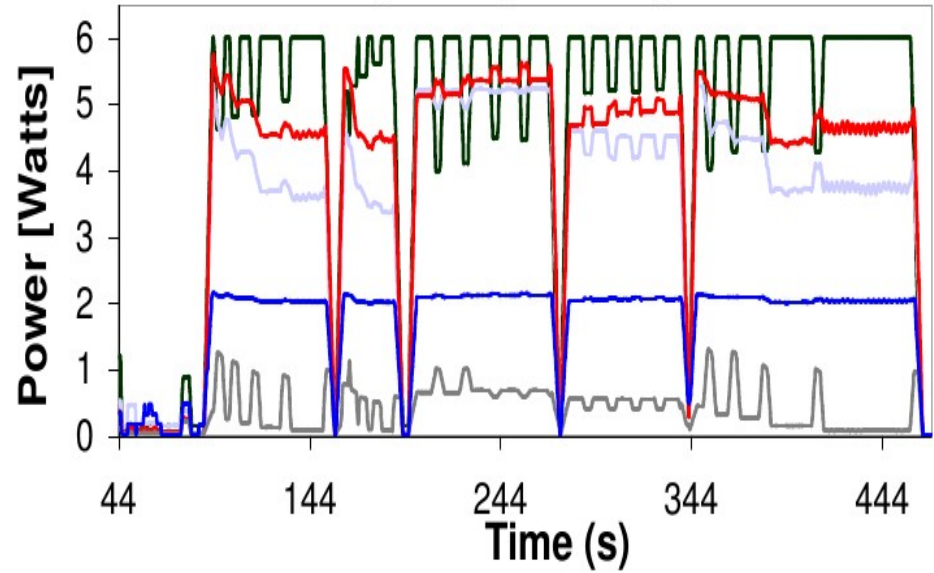
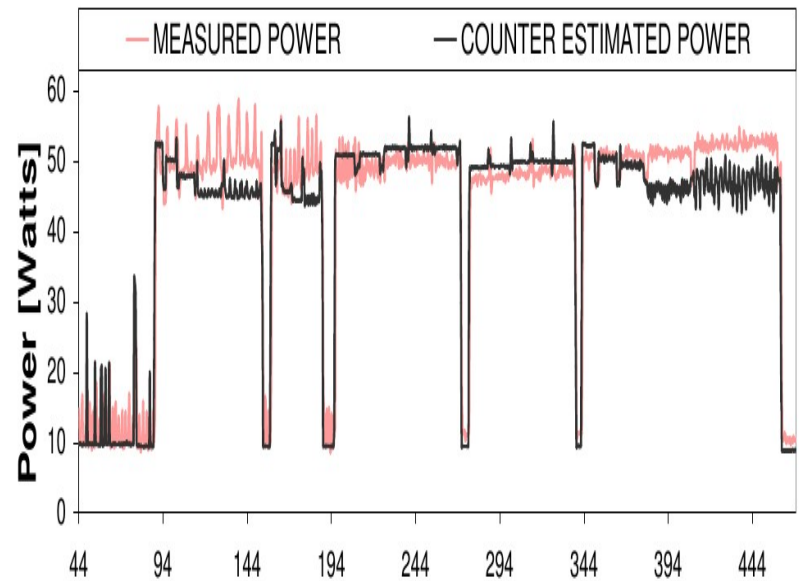


Power vectors for similarity

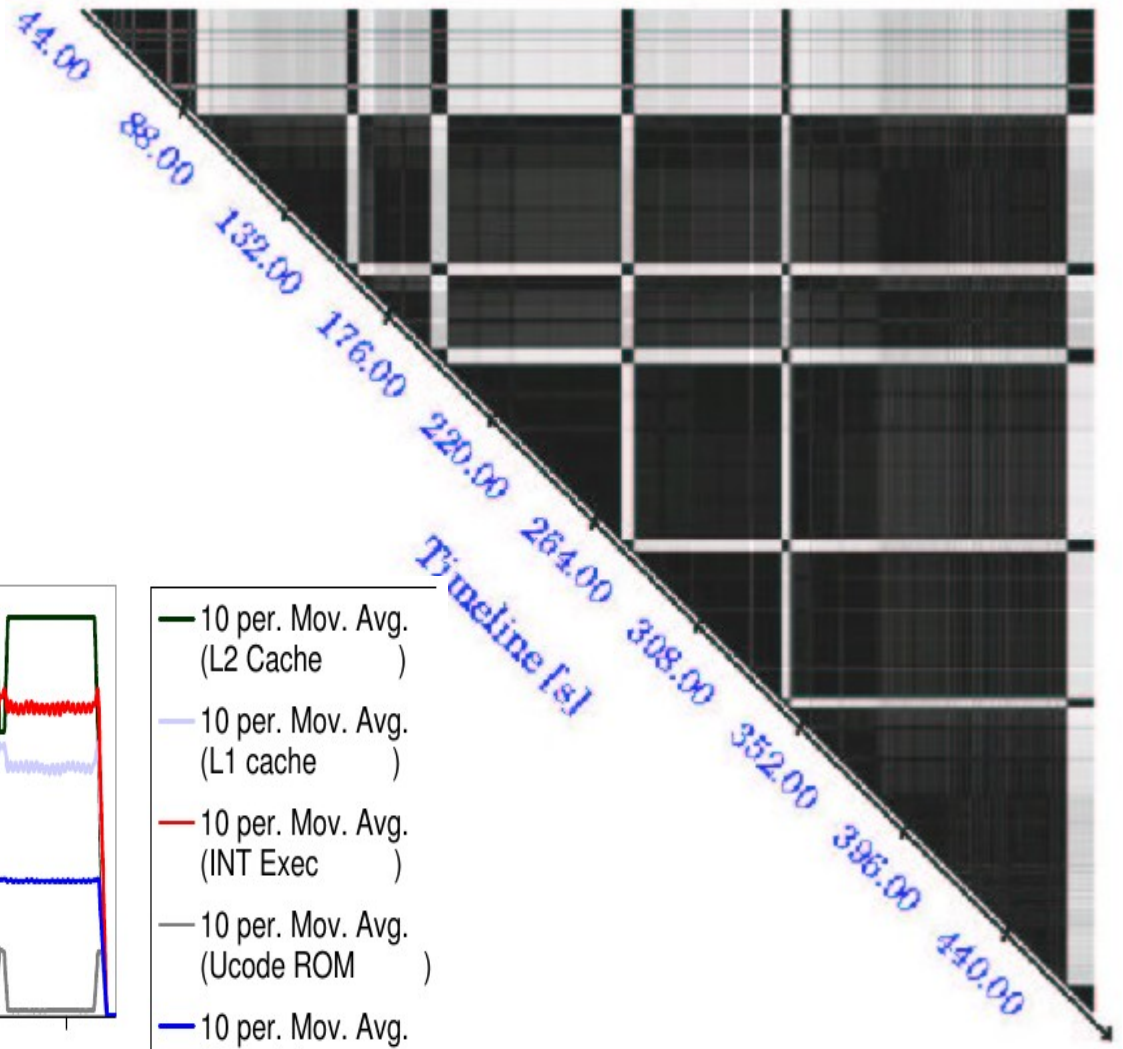
- Spatial closeness of points specified by corresponding power vectors
- Manhattan distance as measure of closeness
- 0 = perfect similarity (black)
- Diagonal = time

First try: based on total power

$$\text{Total Similarity Matrix}(r, c) = |\text{Total Power}_r - \text{Total Power}_c|$$



- 10 per. Mov. Avg. (L2 Cache)
- 10 per. Mov. Avg. (L1 cache)
- 10 per. Mov. Avg. (INT Exec)
- 10 per. Mov. Avg. (Ucode ROM)
- 10 per. Mov. Avg. (Schedule)



Developing a better metric

- Base similarity on the (original) power vectors
- Manhattan distance between vectors

- $$\text{Original Similarity Matrix}(r, c) = \sum_{i=1}^{22} |PV_r(i) - PV_c(i)|$$

- Problem:
 - vectors of smaller magnitude are bound to be considered similar even though they may point to very different directions in power space.

Developing a better metric (2)

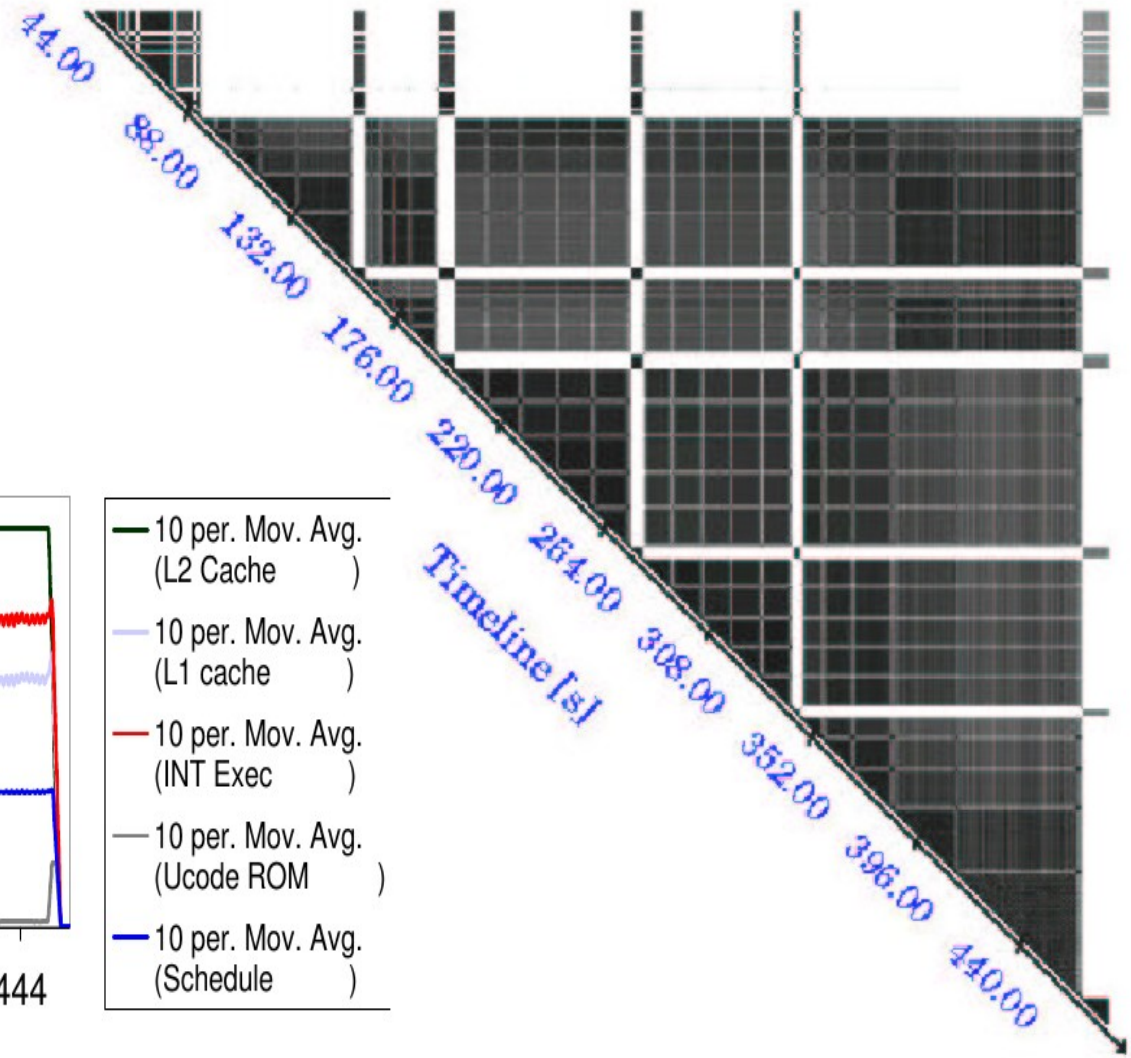
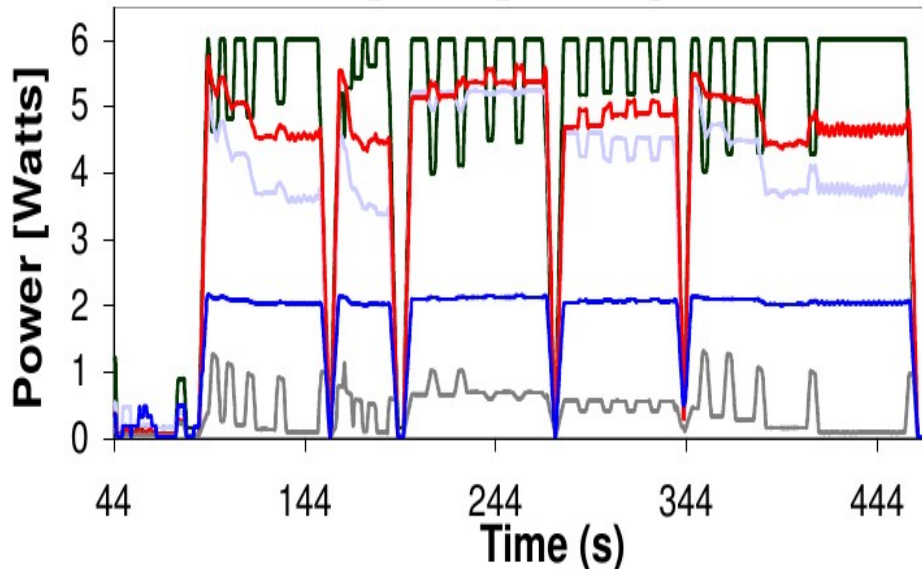
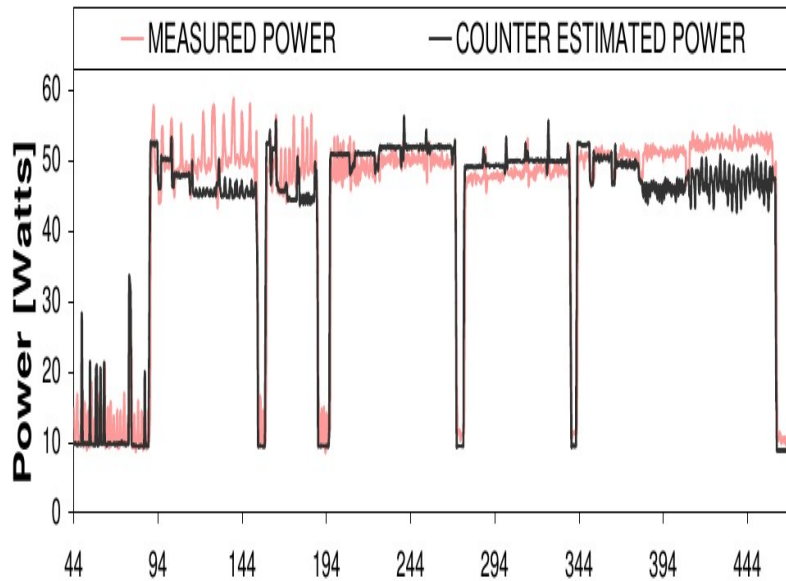
- Normalize vectors to avoid this pitfall

$$\textit{Normalized Similarity Matrix}(r, c) = \sum_{i=1}^{22} |NPV_r(i) - NPV_c(i)|$$

- But: indifferent to magnitude of vectors as ratios prove to be similar
→ combine both metrics!

The final metric

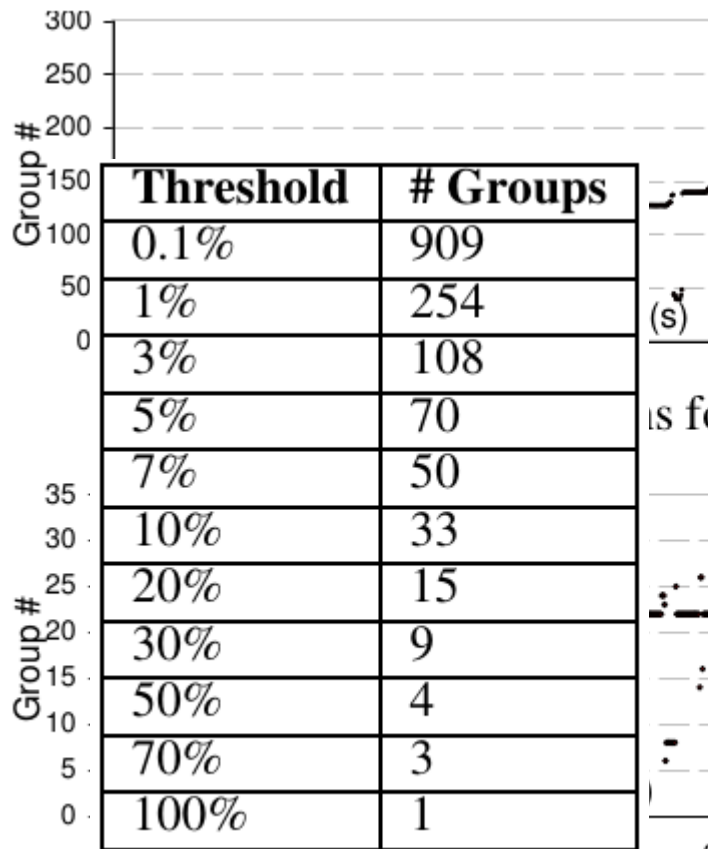
$$FM(r, c) = \min \left(\frac{OM(r, c)}{\max_{r', c'} (OM(r', c'))} + \frac{NM(r, c)}{\max_{r', c'} (NM(r', c'))}, 1 \right)$$



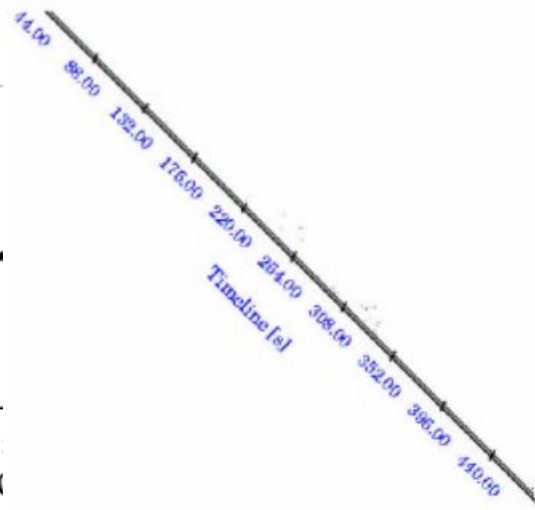
Similarity Groups

- Aim:
 - reduced workload size for benchmarks
 - still capture most of its power behavior
 - power signature using representative vectors
- Goals for Thresholding:
 - (1) Grouping execution points—power vectors—based on their similarity
 - (2) Representing power behavior with reasonable accuracy with a small number of “signature vectors”

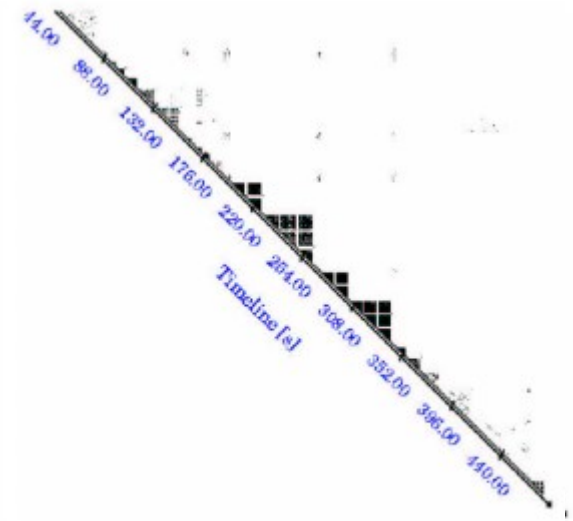
Thresholding



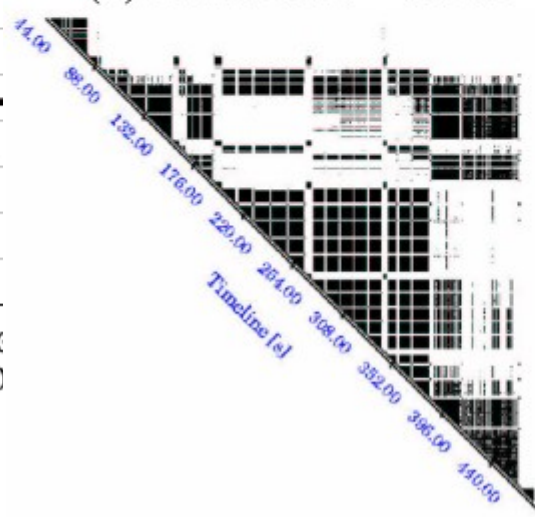
(b) Gzip group distributions for



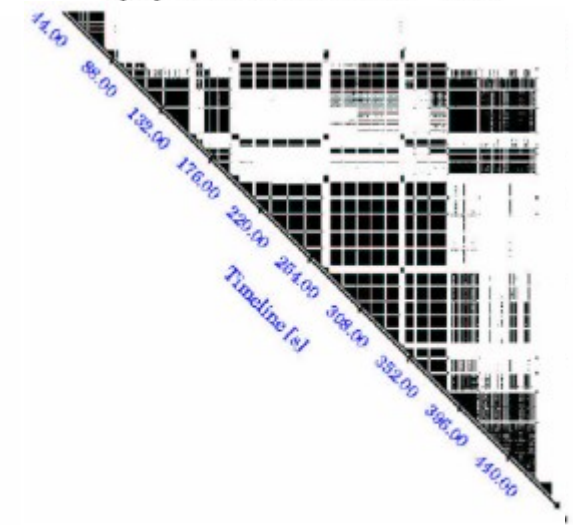
(a) Threshold = 0.1%



(b) Threshold = 1%

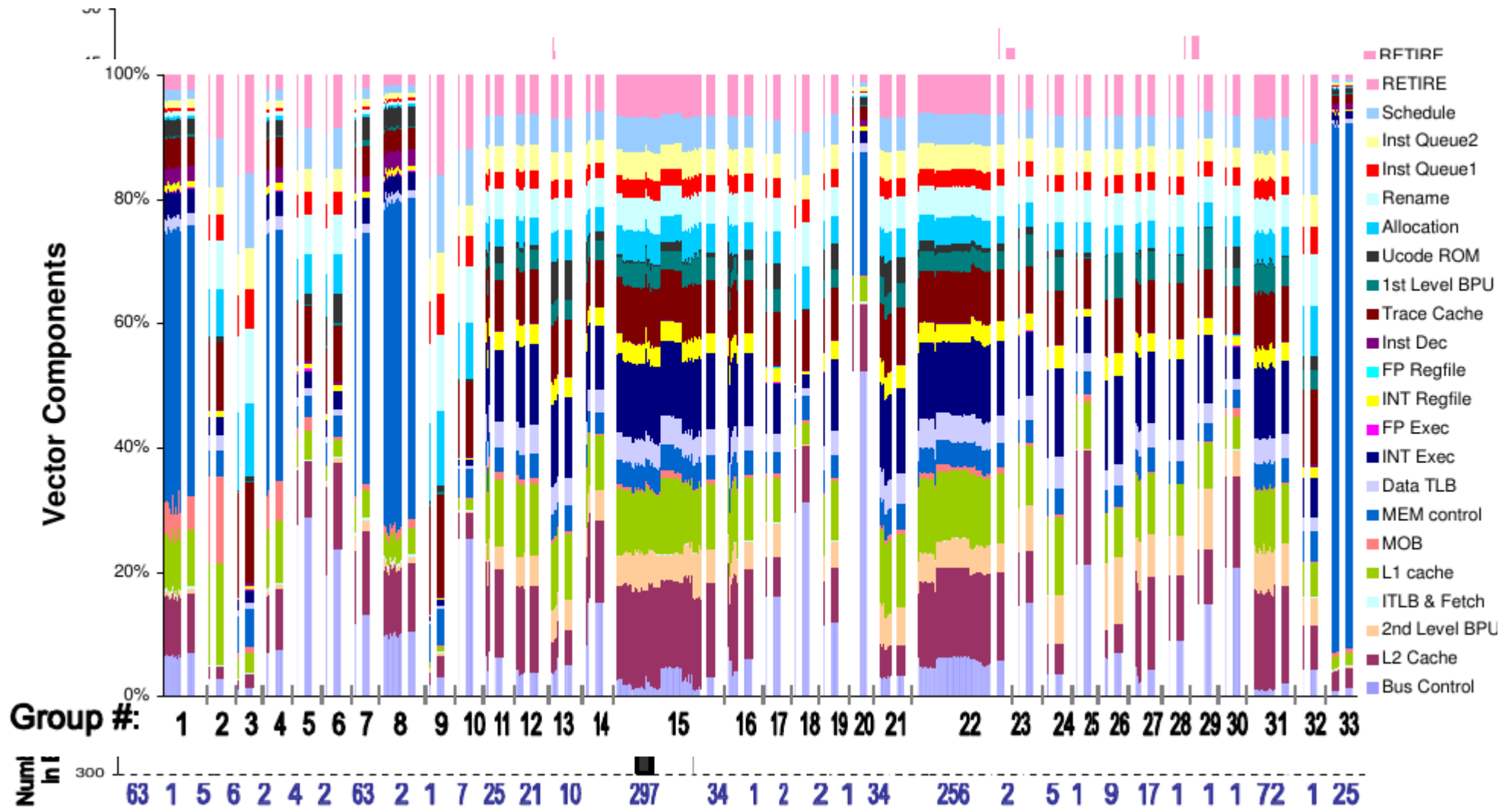


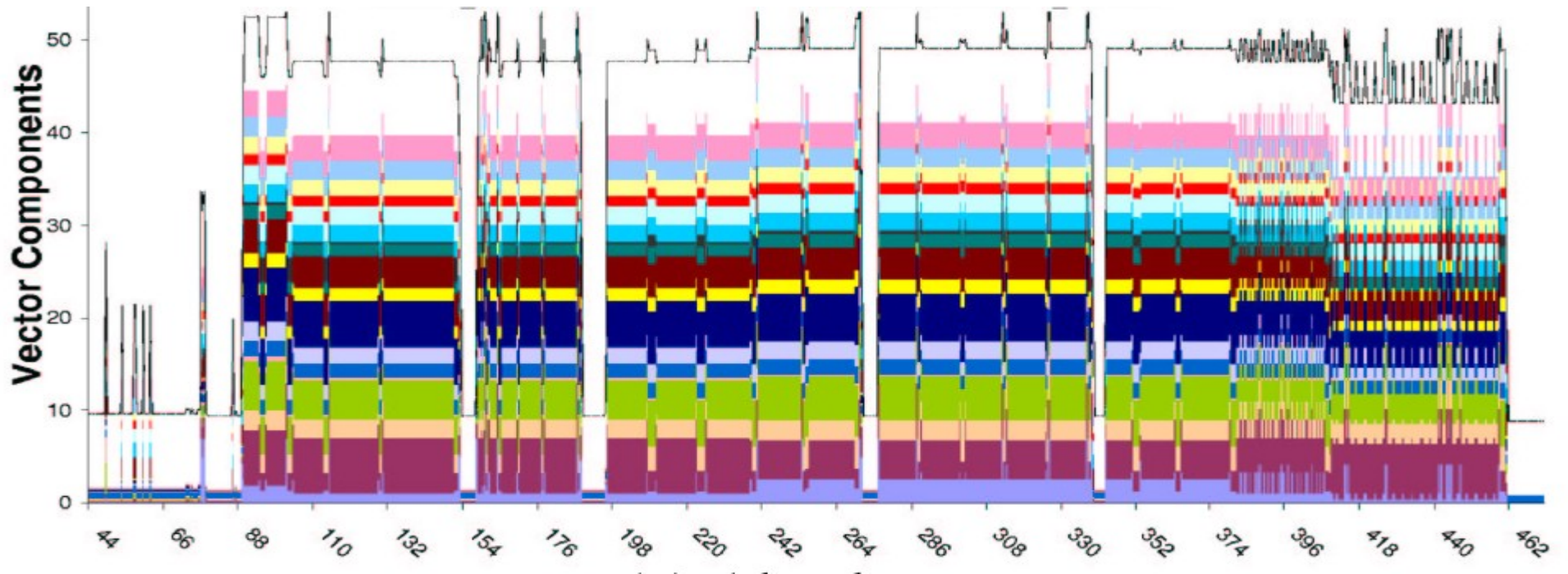
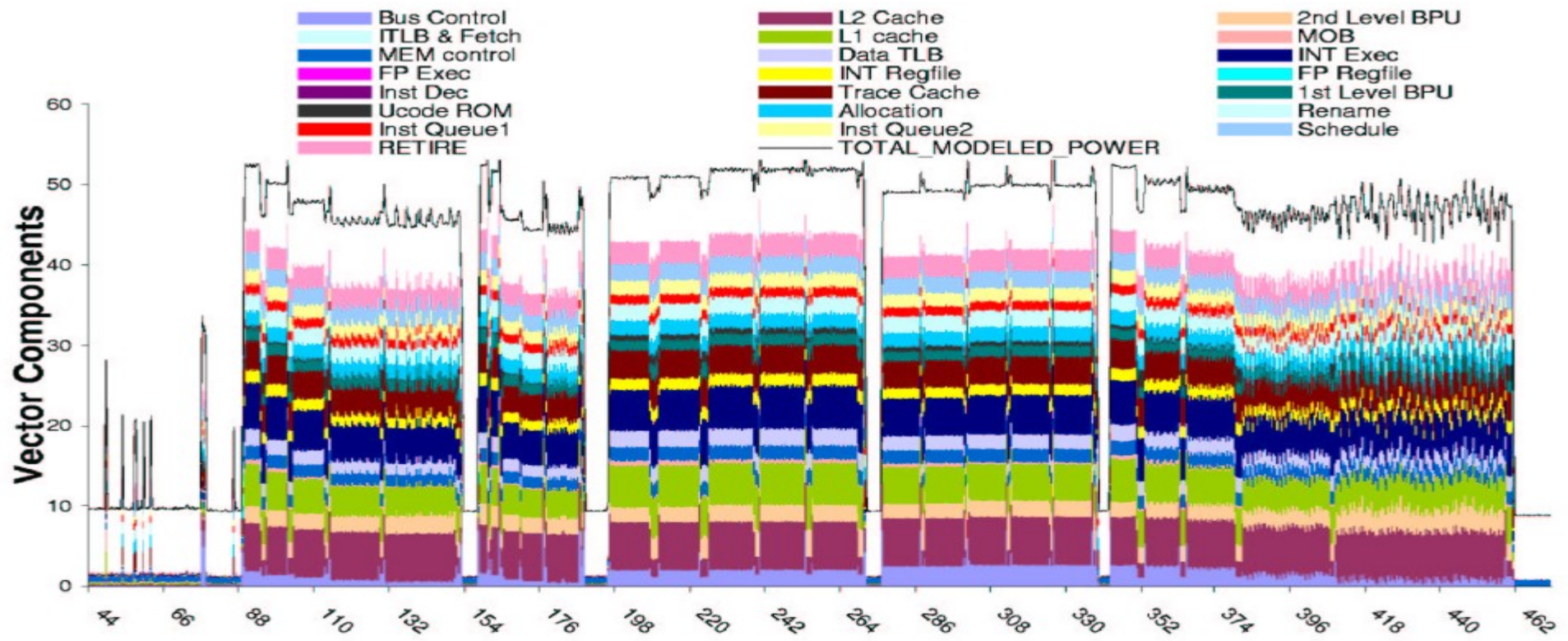
(c) Threshold = 10%



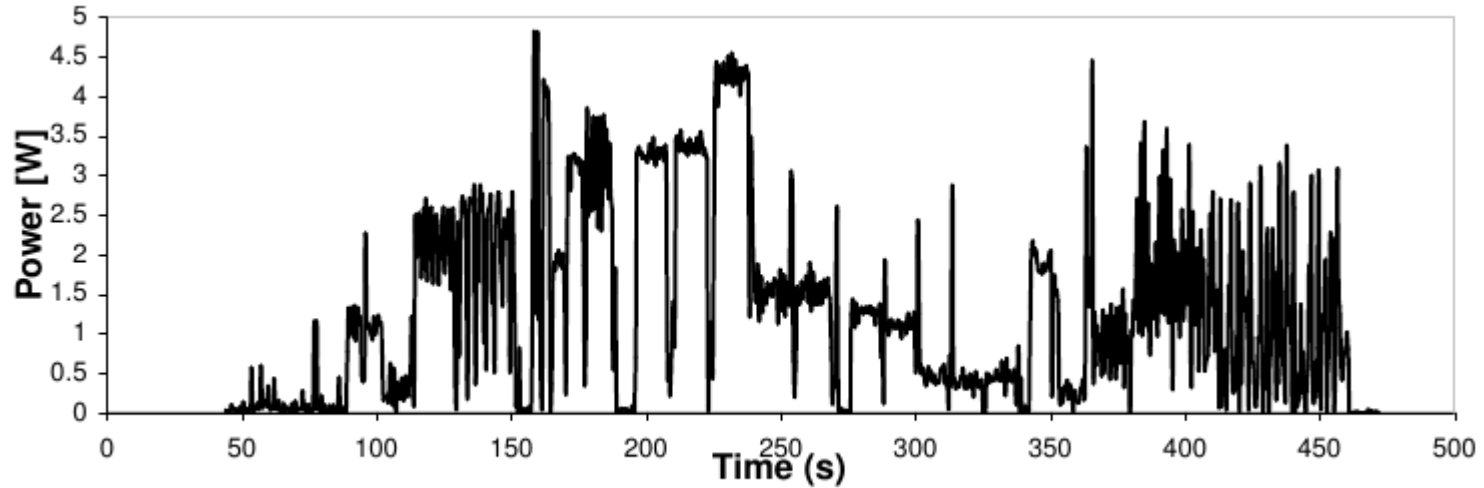
(c) Threshold = 10%

Representative Vectors

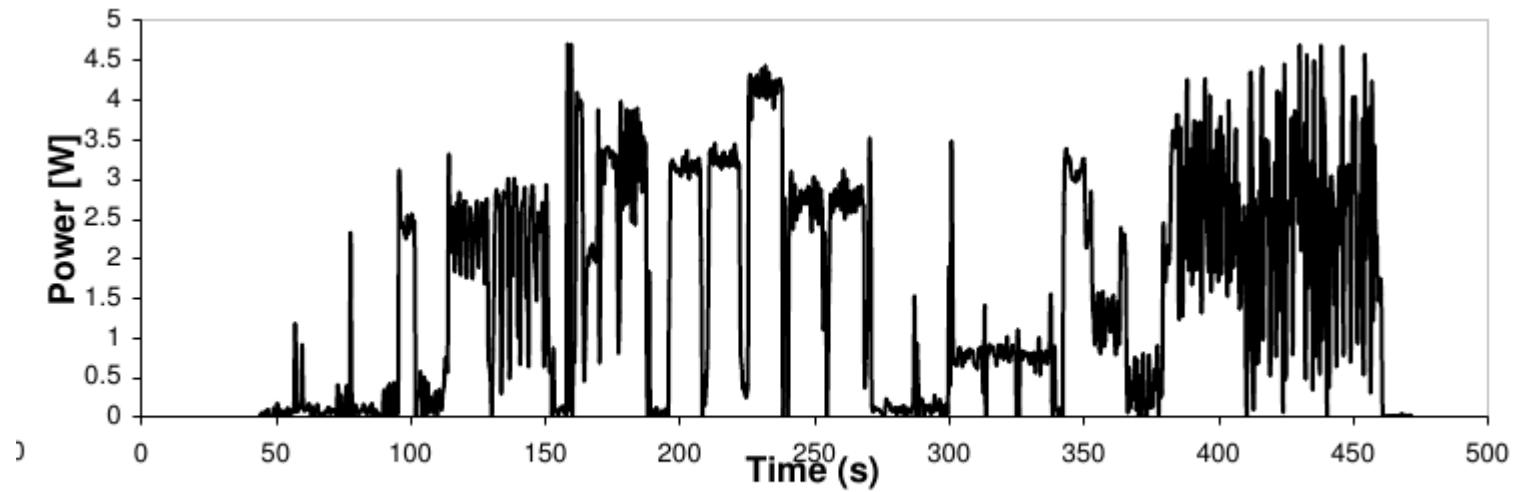




Error Analysis



(a) Error for representative vectors



(b) Error for selected execution points

Conclusion

- Defined combined similarity metric
- Found only considering total power conceals power phase information
- Grouping of vectors based on thresholding
- Generation of signatures based on representative vectors

Discussion

- Only show errors based on simulation :(
- Nice method to reduce benchmarking time :)