

# Similarity Measurement for Proxy Application Fidelity

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### Motivation and Goal

- Quantitatively understand the correspondence of proxy to parent applications with respect to their underlying node and memory hardware behavior.
- Ensure the system co-design and procurement that proxy application are high fidelity models of the target applications
- Find the most efficient similarity measurement.

### Experiment

#### Applications

- ECP Applications in scientific domain of cosmology, seismic modeling, molecular dynamics, and thermal transport.
- 16 apps in proxy/parent pairs and 5 unpaired.

#### System platform

- Intel Skylake
- IBM Power9

MPI-only mode, 128 ranks, 1 rank/core, on four nodes.

#### Data collection

- Collection infrastructure: LDMS
- PAPI sampler collect 700 hardware events performance for each application.
- Run same problem for each application:
  - Use the same input problem and/or parameters.

### Methodology

- 1. Application representation
- A vector with 700 dimensions (features), each of which is an average rank value for one hardware event.
- 2. Feature set
  - Overall (700 features)
  - Subset (subgroup features):
    - Cache-related, e.g., longest\_lat\_cache.miss
    - Instruction mix
    - Pipeline
- 3. Distance measurement
  - Euclidean distance:
    - Cosine similarity

$$\cos\theta = \frac{\sum_{i=1}^{n} x_i y_i}{\|x\| \|y\|}$$

- Statistic distance:
- Kullback–Leibler divergence

$$D_{\mathrm{KL}}(P \parallel Q) = \sum_{x \in \mathcal{X}} P(x) \log igg(rac{P(x)}{Q(x)}igg)$$

Jensen-Shannon divergence

$$\mathrm{JSD}(P \parallel Q) = rac{1}{2} D(P \parallel M) + rac{1}{2} D(Q \parallel M)$$

Wasserstein distance

$$W_2(p,q) = \sqrt{\min_{P_{XY}} E_{P_{XY}} \left[ \|x-y\|_2^2 
ight]} ext{ s. t. } P_X \sim p, P_Y \sim q$$

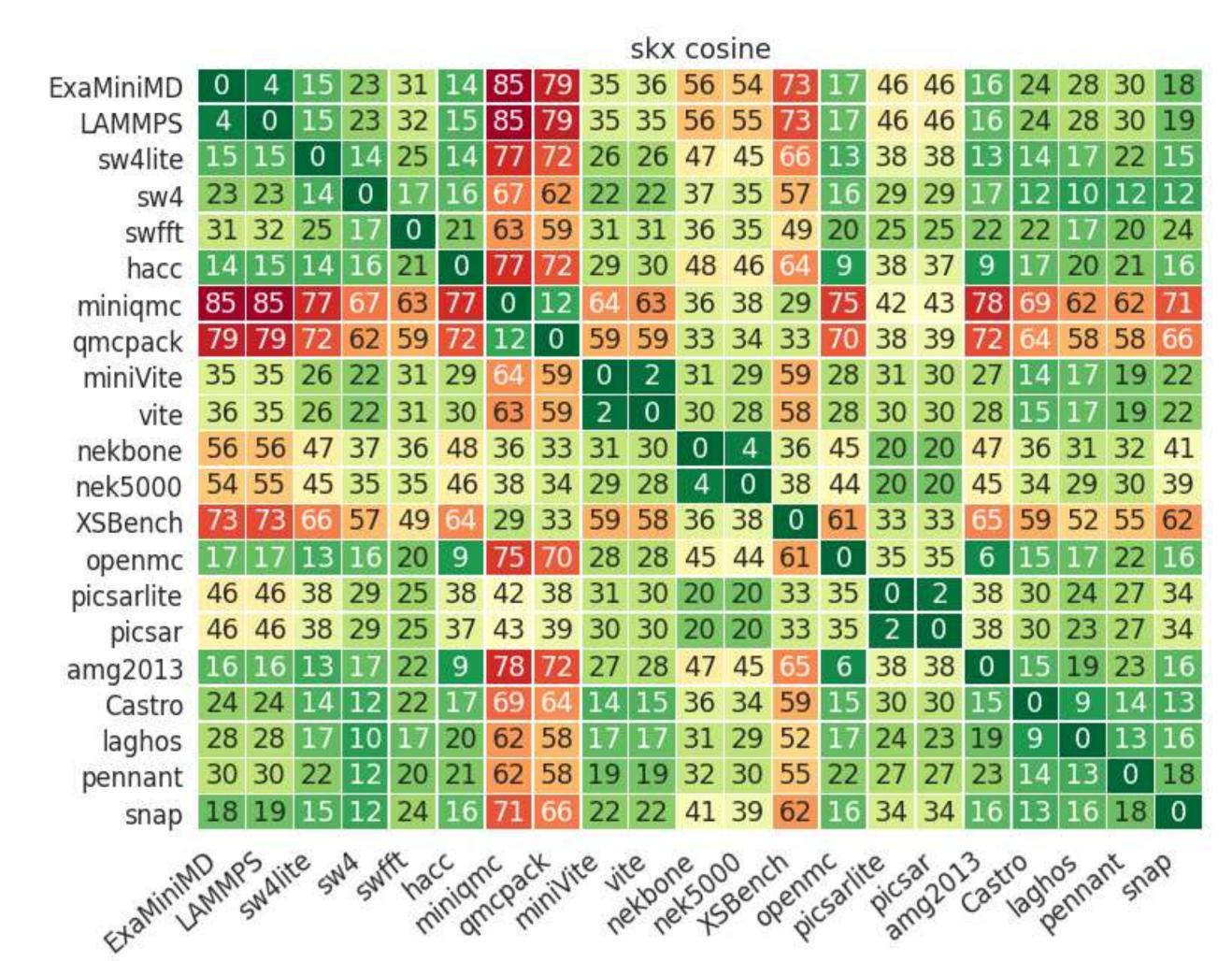
- 4. Cluster similar applications
  - Hierarchical Clustering
    - Get the similarity by tree branch height to rank applications.

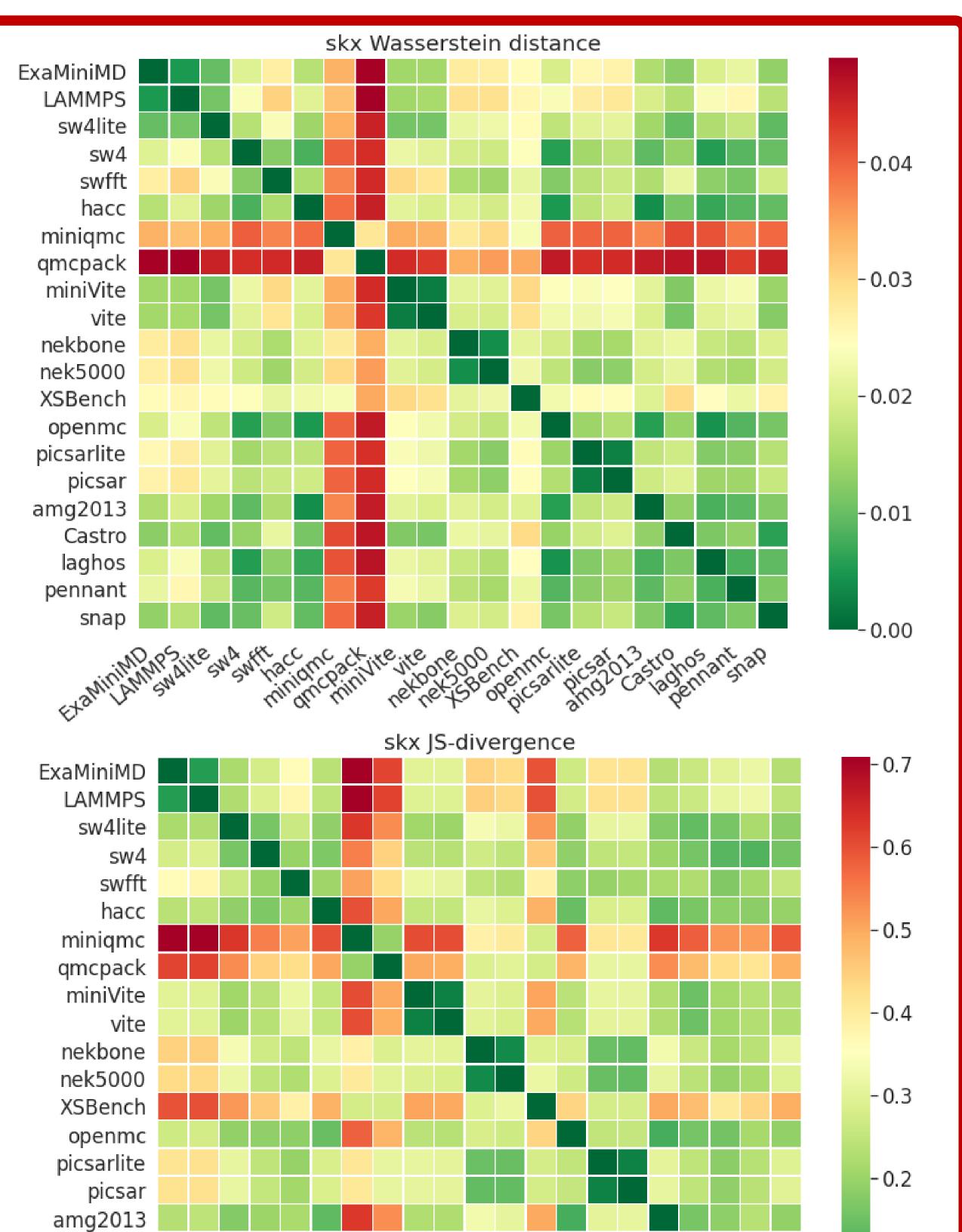
**ACKNOWLEDGEMENTS** 

- 5. Evaluation
  - Similarity ranking
  - Self similarity

## Result

- Overall feature results of three distance methods in the Skylake system look alike.
- Three apps are generally different from the others.
- Three pairs are the most similar proxy/parent pairs.
- Two pairs (nek5000/nekbone and vite/miniVite) show high similar performance in cosine similarity, and relatively high similar performance in other two methods.
- The last five unpaired applications show relative similarity.
- QMCPack and MiniQMC are similar but differ from other applications.





### Conclusion

- Overall ranking show similar result for three distance methods.
- Cross platform show different result because of different memory subsystem and SIMD width.
- Subset results: three subgroups contribute most to the overall similarity matrix.
- Most proxies are good representations of their parents, only with few divergences.
- Cosine similarity is simple, quick and interpretable by geometric angle
- Since other distance methods show similar result, cosine similarity is good enough to use for similarity measurement.

### Future work

Castro

- Include domain expert evaluation for the choice of distance method.
- Calculate self-similarity for time series data with sliding windows to further validate the similarity methods.
- Validate similarity through measures of latency or some other environmental / real metric.