

# **EFFECTIVE KERNEL MAPPING FOR OPENCL APPLICATIONS IN HETEROGENEOUS PLATFORMS**

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# OUTLINE

- Introduction
- System Overview
- Our Approach
- Results
- Future Work
- Conclusion

# INTRODUCTION

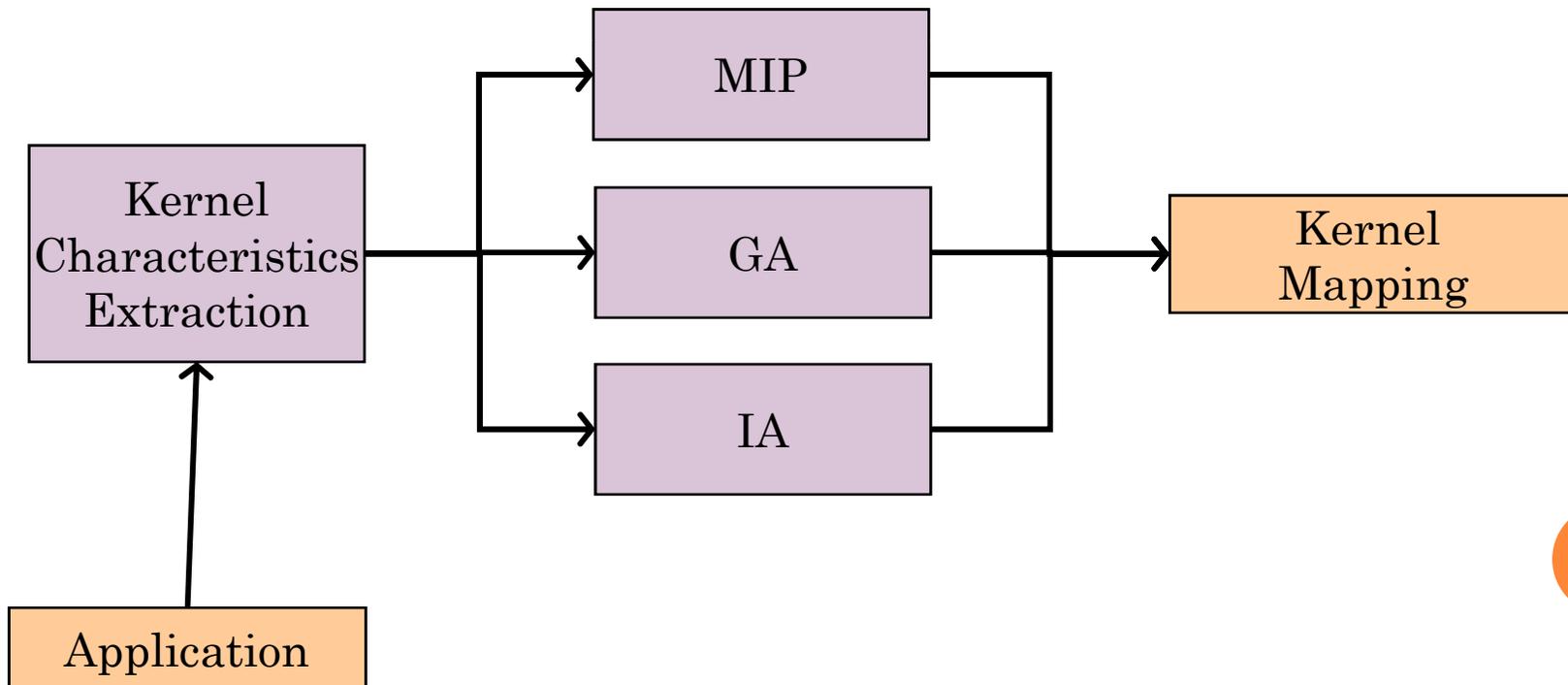
- Parallel computing systems consist of different types of accelerators
- Applications that are running on these systems have also different needs
  - Processing
  - Memory
  - Communication
  - Storage
- Use resources effectively
  - Application performance
  - Resource utilization

# INTRODUCTION

- Applications composed of multiple exclusive regions
  - Kernels
  - Variation in characteristics and requirements
- Our Approach
  - Profiling-based kernel mapping algorithm
  - Multi-kernel applications
  - Heterogeneous platforms

# SYSTEM OVERVIEW

- Collect statistics about the kernels
- Select the best device for each kernel
- Execute the kernel on the chosen device

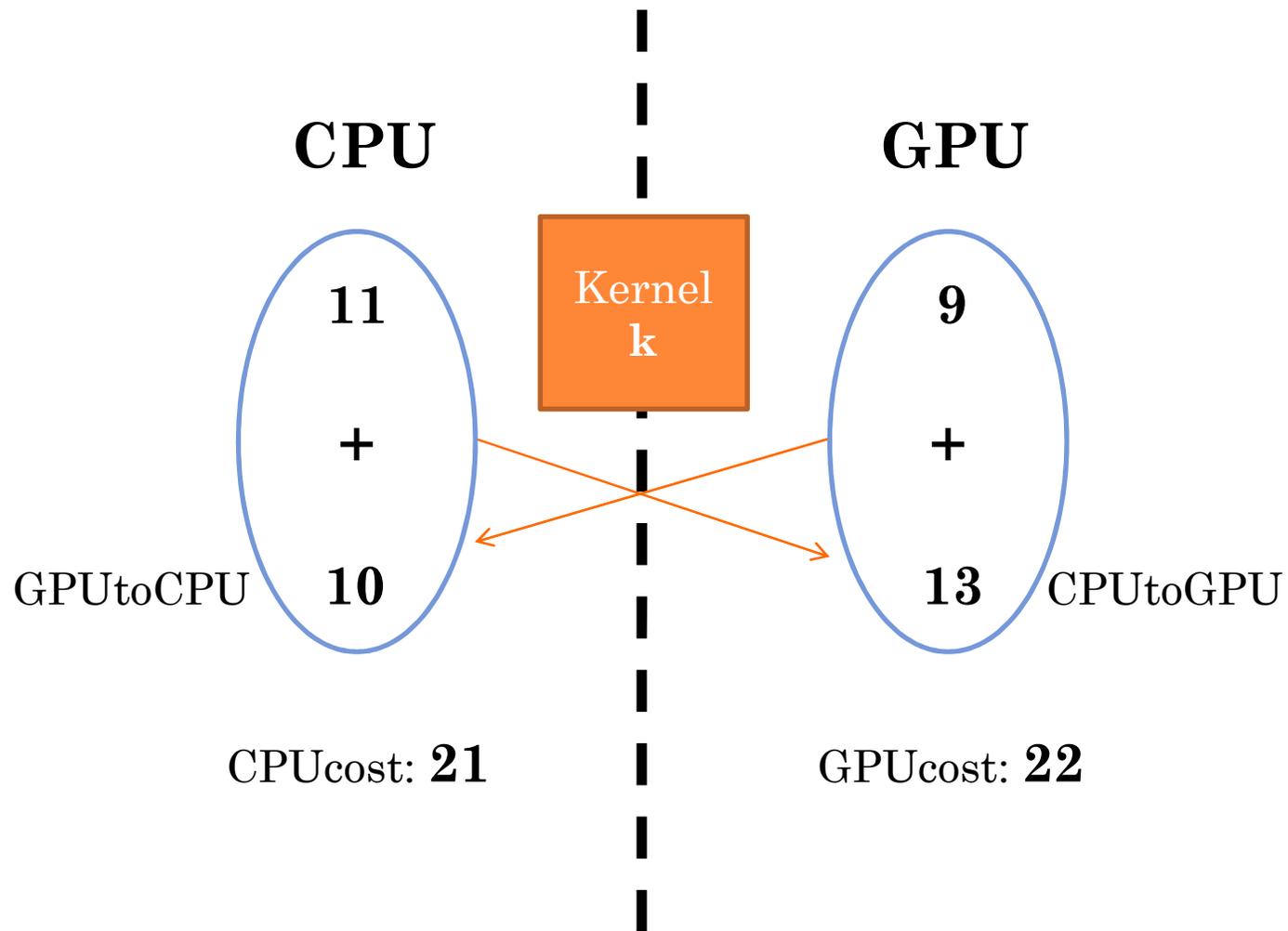


## GREEDY ALGORITHM

- Find the best device for each kernel starting from the first one

$$\begin{aligned} CPUcost_k &= CPUrunningtime_k + \\ &\sum_{d=1}^n DeviceToHost \times InDevice_d \times \\ &\quad Required_{k,d} \times size_d. \end{aligned}$$

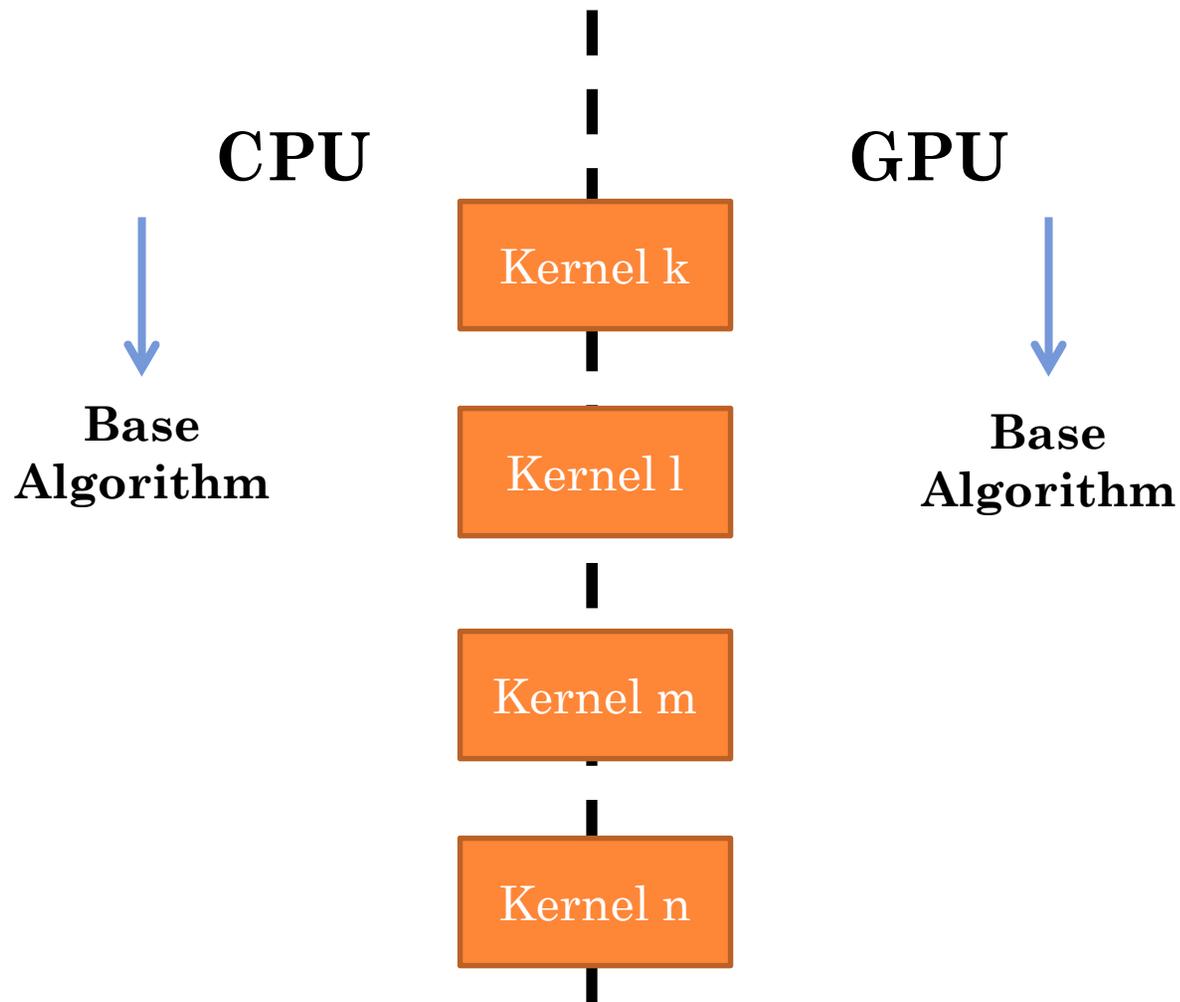
# GREEDY ALGORITHM



# IMPROVED ALGORITHM

- For each kernel, force to explore both CPU and GPU
  - Run Base Algorithm by taking CPU path
  - Run Base Algorithm by taking GPU path
  - Compare the overall performances
  - Execute kernel k on the corresponding device

# IMPROVED ALGORITHM



# EXPERIMENTAL SETUP

## ○ System

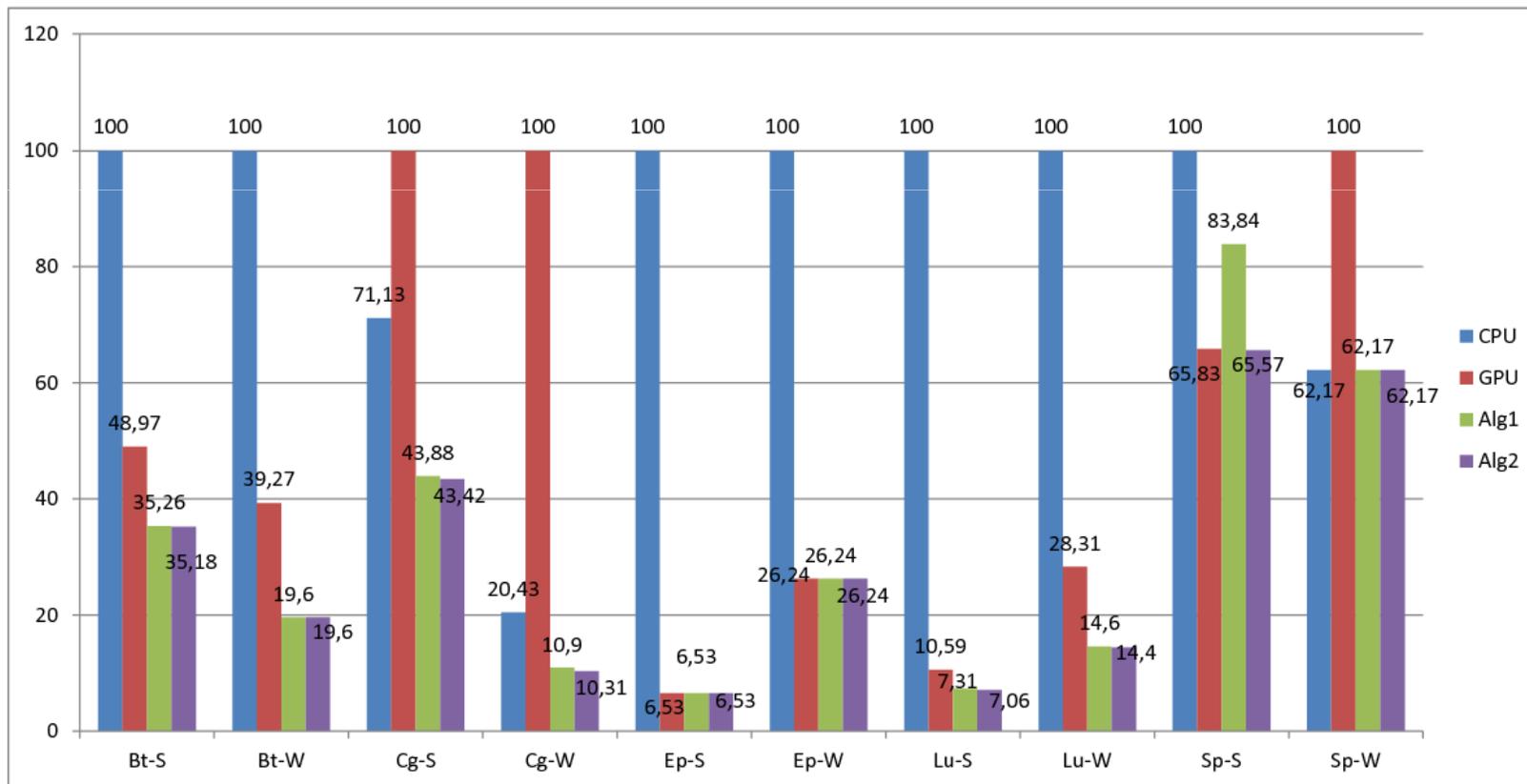
CPU	GPU
AMD Phenom II X6 1055T	GeForce GTX 460
AMD APP SDK v2.6	NVIDIA OpenCL SDK 4.0
4GB	1GB

## ○ NAS Benchmarks [1]

Bench.	Description
BT	Solves non-diagonally dominant, block-tridiagonal equations.
CG	Computes an approximation for sparse, symmetrically positive definite matrix using a conjugate gradient method.
EP	Evaluates an integral by means of pseudo random trials.
LU	A regular-sparse, lower/upper triangular system solution.
SP	Solves non-diagonally dominant, pentadiagonal equations.

# EXPERIMENTAL RESULTS

- Except for EP-S, EP-W, and SP-W, all benchmarks use both CPU and GPU resources



# RELATED WORK

- Profiling

- Luk et al.[2]

**Qilin:** predict kernel execution times offline

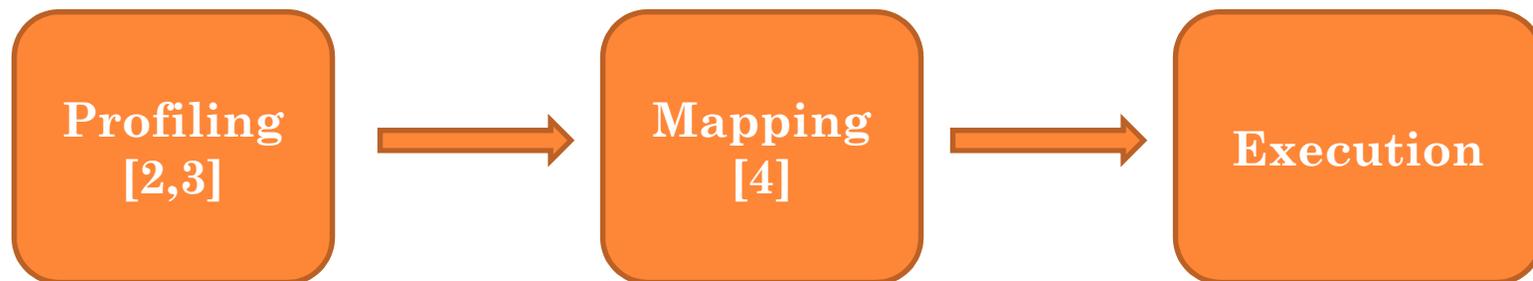
- Grewe et al.[3]

A static machine learning-based task mapping algorithm

- Mapping

- Augonnet et al.[4]

**Starpu:** Runtime system tool that provides an interface unifying execution on accelerator technologies as well as multicore processors.



## FUTURE WORK

- Memory controller is not considered in current implementation
- Check for independent kernels and allocate them onto devices in a parallel fashion
- Use multiple GPUs

# CONCLUSION

- Effective kernel mapping for heterogeneous systems
  - Using CPUs and GPUs
  - Heuristic
  - ILP
- Our algorithm effectively assigns the kernels to the devices

## REFERENCES

- [1]“NASA, NAS parallel benchmarks  
<http://www.nas.nasa.gov/publications/npb.html>
- [2] C.-K. Luk, S. Hong, and H. Kim, “Qilin: exploiting parallelism on heterogeneous multiprocessors with adaptive mapping,”
- [3] D. Grewe and M. F. P. O’Boyle, “A static task partitioning approach for heterogeneous systems using opencl,”
- [4] C. Augonnet, S. Thibault, R. Namyst, and P.-A. Wacrenier, “Starpu: a unified platform for task scheduling on heterogeneous multicore architectures,”

Thanks!