

**INTRODUCTION**

- Linear Programming (LP) models are frequently encountered optimization problems in medicine, engineering and business, where a linear objective function needs to be either minimized or maximized subject to multiple linear constraints.
- High performance implementations of the Simplex Algorithm facilitate solutions to large problems in these disciplines.
- This work proposes an energy-efficient hardware accelerated LP solver for a class of problems in radiation therapy. The proposed accelerator is implemented and tested on various devices using OpenCL.

**LINEAR PROGRAMMING**

- The standard form of a linear programming problem is (1), where  $A$  represents the constraint system,  $b$  represents the constraint bounds,  $c$  is the cost function, and  $x$  is the decision variables.

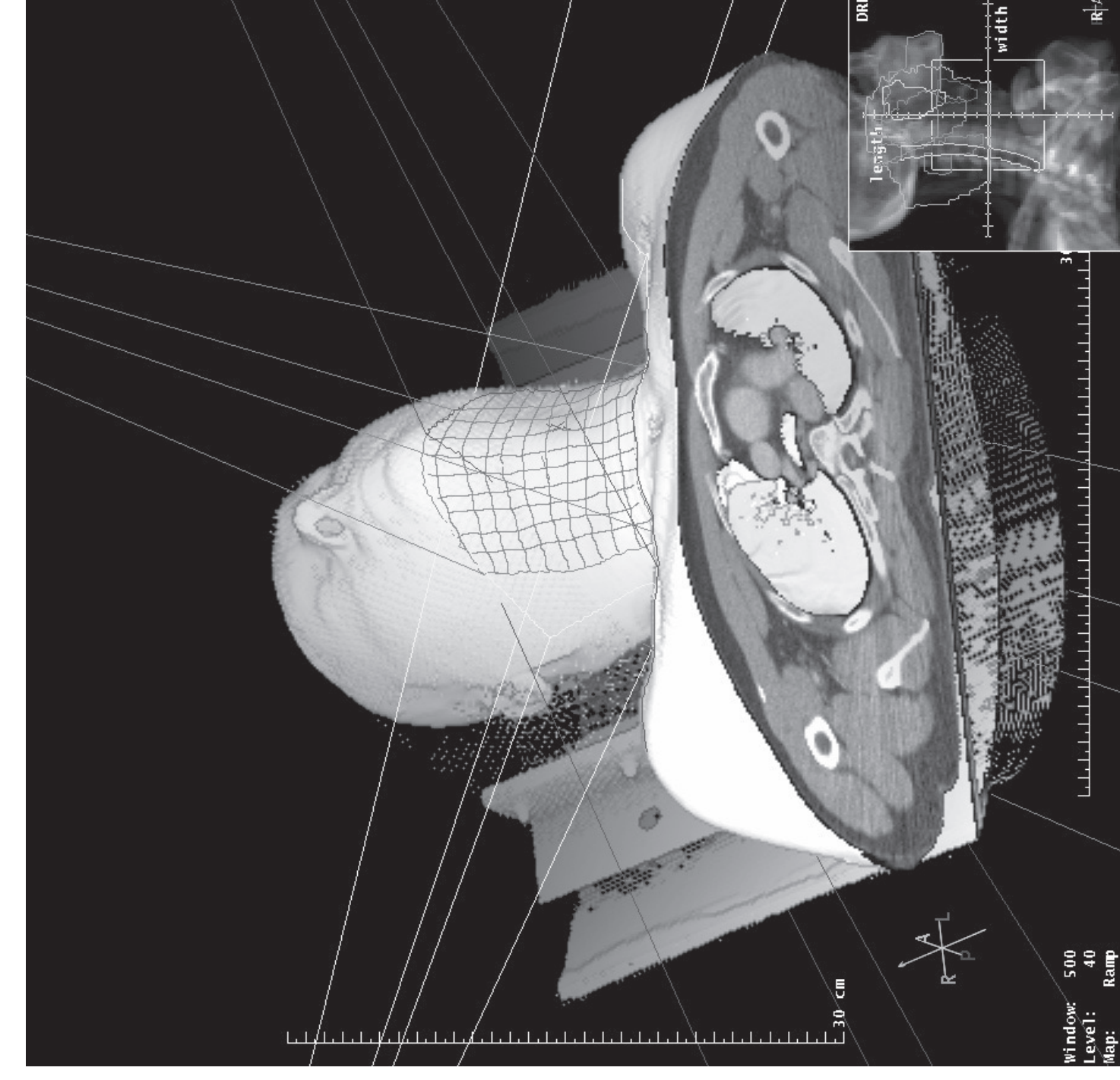
$$\begin{aligned} & \min \quad c^T x \\ & \text{s.t.} \quad Ax \leq b \\ & \quad \quad x \geq 0 \end{aligned} \quad (1)$$

**APPLICATION: RADIATION THERAPY TREATMENT PLANNING**

- Radiation therapy uses several beams to deliver a radiation dose to tumour cells. Recent literature has formulated the problem of optimizing the beam weights as an LP problem [3-4].
- The affected area is divided into target  $t$  and normal  $n$  voxels. The radiation from each beam to each voxel is measured to form the dose matrix  $D$ .
- The model (2) minimizes the total dose delivered to non-target areas subject to constraints based on the properties of the tissue in each voxel.

$$\begin{aligned} & \min \quad (D_n^T c_n)^T w \\ & D_t w \leq x_t^U \\ & D_t w \geq x_t^L \\ & D_n w \leq x_n^U \\ & w \geq 0 \end{aligned} \quad (2)$$

Figure 1: Radiation Therapy Beam Configuration [4]



- Vectors  $x$  represent the upper  $U$  and lower  $L$  bounds for the target and normal areas. Beam weights  $w$  are given costs  $c_n$ .
- Since each beam will deliver radiation to most voxels, the dose matrix of the LP problem will be dense.

**OPENCL**

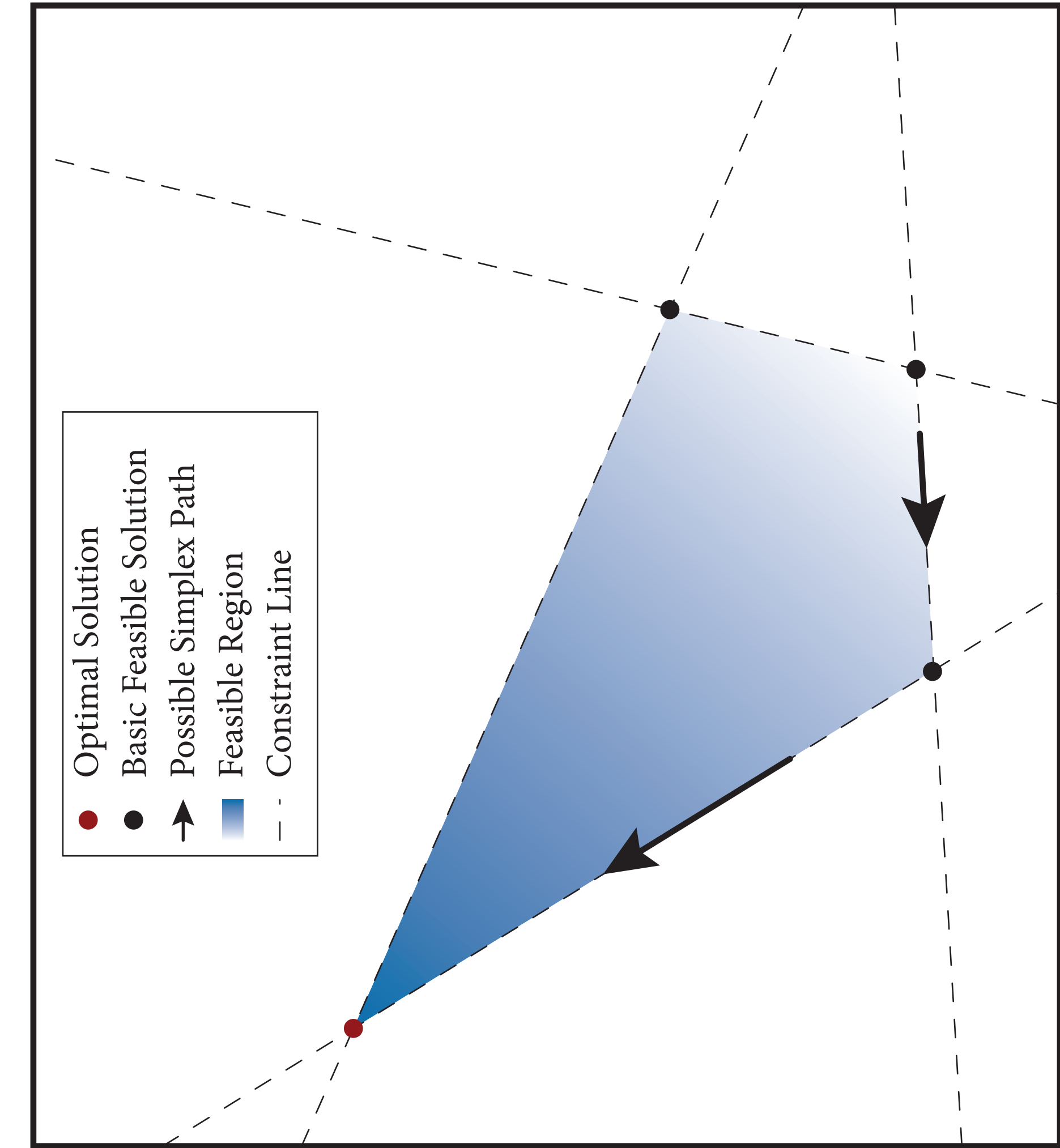
- Language designed by Khronos Group for portability across hardware accelerators.
- A Hardware accelerator, or OpenCL Device, is managed by a Host processor.
- The OpenCL linear programming engine was tested on multiple devices to measure acceleration (see Table 1 for specifications).
- Effort was made to optimize OpenCL kernels for each device without losing design portability.

Table 1: OpenCL Test Device Specifications

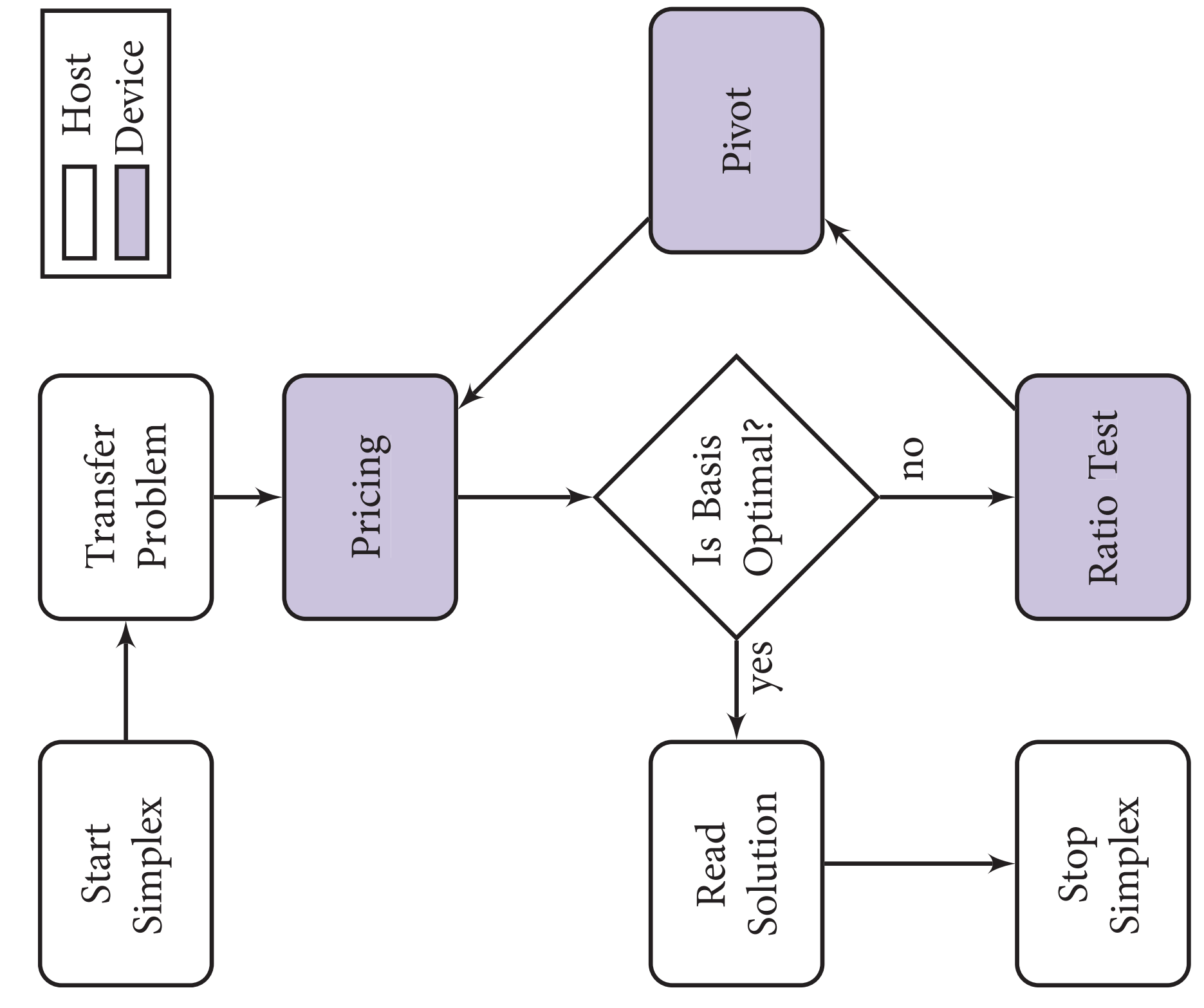
Device	Power (W)	Memory Bandwidth (GB/s)
Intel Core i7 4930k (6 cores)	130	59.7
Nvidia GeForce GTX-780	250	288.4
Altera Nallatech PCIe-385N	25	24.0

**THE SIMPLEX ALGORITHM**

- The Simplex Algorithm solves LP problems by iterating over subsets of the problem variables to generate improving solutions.
- Each simplex iteration consists of three subroutines: pricing, ratio test, and pivot.
- Pivot finds the direction of the adjacent solution, ratio test calculates the magnitude of the reduction in the cost function, and pivot updates the problem data structures.
- The three subroutines are performed on the OpenCL Device in a loop managed by the Host processor.



a) The Simplex Algorithm



b) OpenCL Implementation

**PERFORMANCE BENCHMARKING**

Figure 3: Speed Up for the OpenCL Solver Over the Sequential Solver

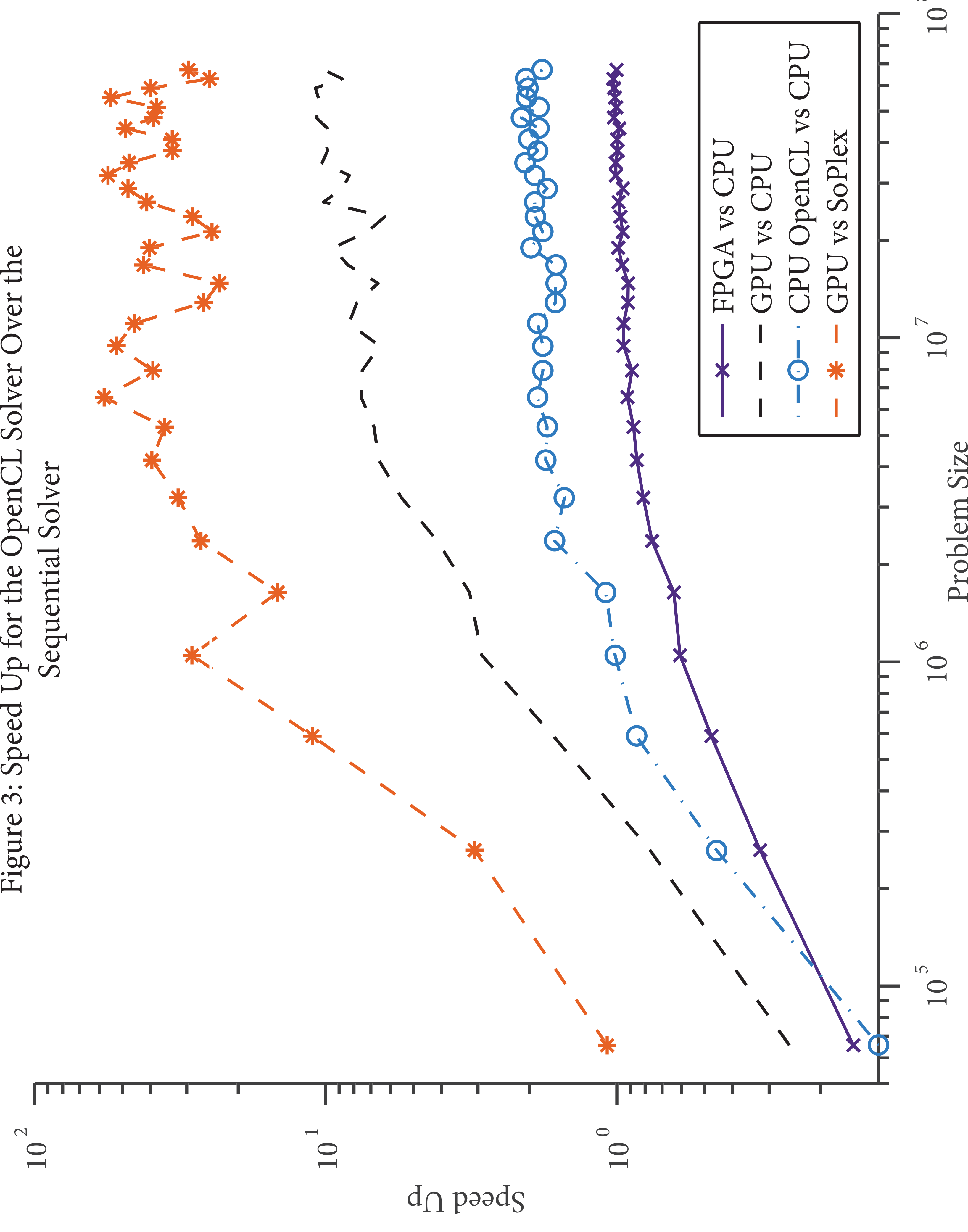
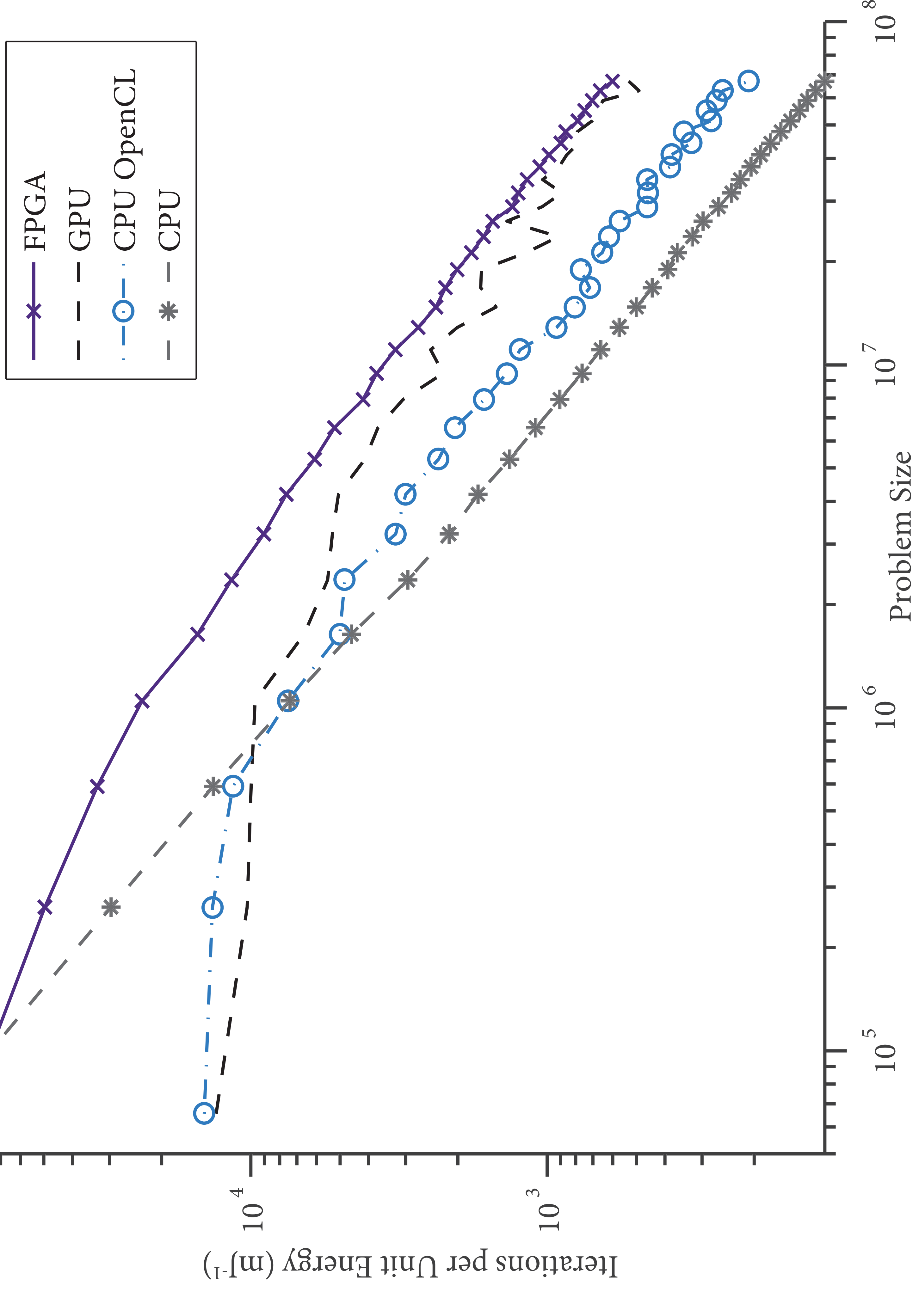


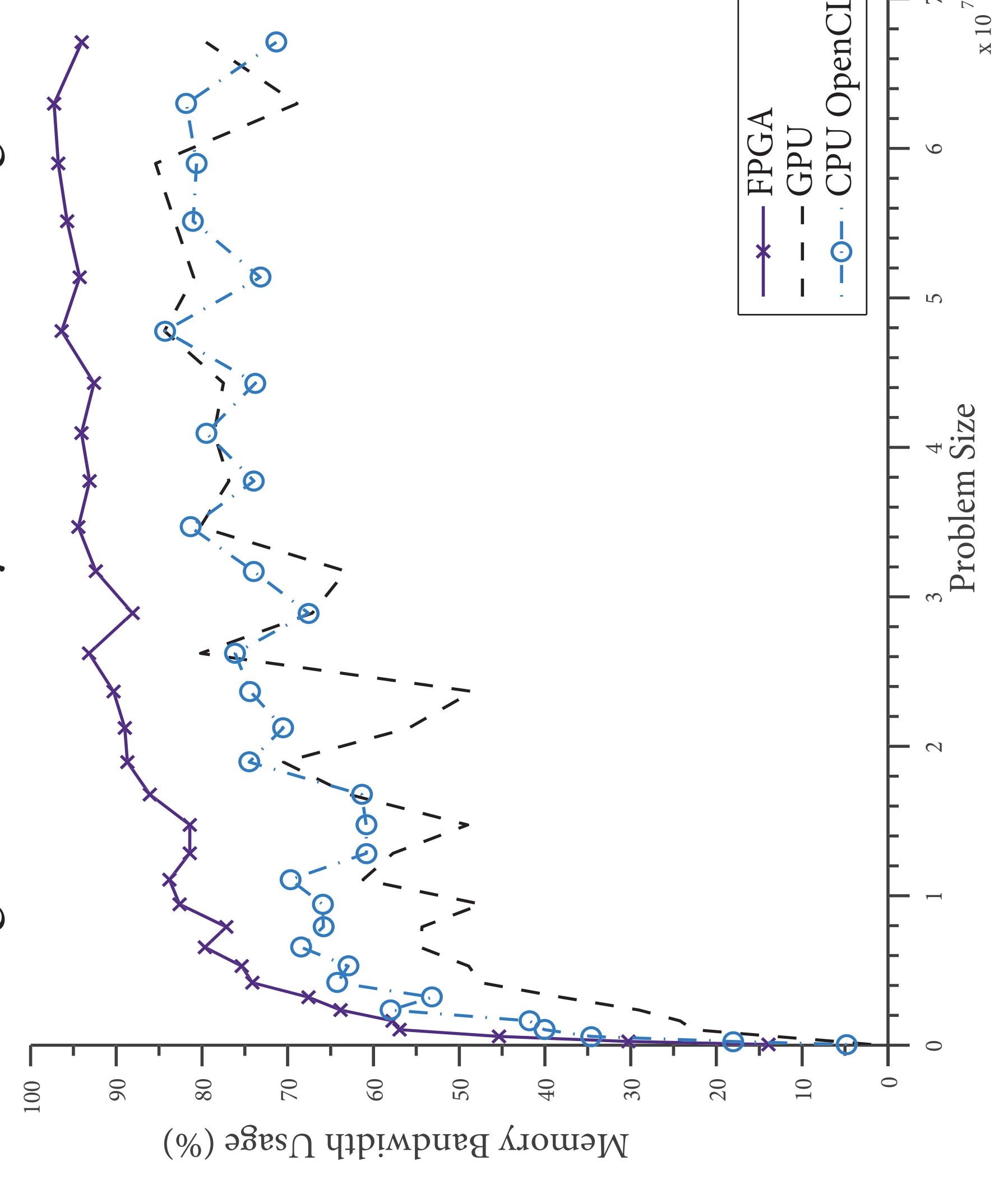
Figure 4: Energy Efficiency of the OpenCL and Sequential Solvers



**DESIGN ANALYSIS**

- Performance benchmarking reveals speed ups relative to sequential code that approach 2 and 10 on a CPU and GPU.
- The FPGA exhibits close to unity speed up but proved to be the most efficient in terms of Simplex iterations processed per unit energy with an efficiency 5 times greater than the CPU.
- The design effectively saturates the memory bandwidth of each test device as shown in Figure 5.

Figure 5: Memory Bandwidth Usage



**CONCLUSIONS**

- A heterogeneous OpenCL system is proposed to solve dense LP problems and tested on multiple hardware accelerators.
- Test results indicate that the GPU provides the fastest solutions to LP problems as shown in Figure 3.
- The selected FPGA provides superior energy efficiency for the application as shown in Figure 4.

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# HARDWARE ACCELERATED LINEAR PROGRAMMING

## Parallelizing the Simplex Method with OpenCL

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