Enabling GPU support for the COMPSs-Mobile framework

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COMPSs-Mobile infrastructure

Cloud

WAN

Collects data from sensors

Offloads computation to Cloud

LAN

PAN

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Programming Model: COMPSs

- Programming
  - Sequential code
  - Standard programming languages with no APIs
  - Infrastructure agnostic
  - Task-based
    - An annotated interface contains the declaration of those methods (CE) selected to become tasks with information about the accesses to data

- Execution managed by a runtime system
  - Replaces CE invocations by asynchronous tasks
  - Monitors data-dependencies among tasks
  - Orchestrates the execution on the underlying infrastructure
    - Transfers data values
    - Submits task executions

Programming Model Example

Matmul.java

```java
public class Matmul {
    public static void main(String[] args) {
        int[][] A;
        int[][] B
        int[][] C;
         ...
        C = multiply(A, B);
        ...
    }
    public static int[][] multiply (int[][] A, int[][] B) {
        // Matrix multiplication code
        //C = AB
        ...
        return C;
    }
}
```

CEI.java

```java
public interface CEI {
    @Method(declaringClass="es.bsc.compss.matmul.Matmul")
    int[][] multiply (
        @Parameter(direction = IN)
        int[][] A,
        @Parameter(direction = IN)
        int[][] B
    );
}
```

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Programming Model Extension

- Add the OpenCL kernel codes in application resource files

- Indicate on the CEI the CE implementation as an OpenCL kernel using the **@OpenCL** annotation indicating:
  - Resource containing the code implementation (kernel)
  - Number of work-items (globalWorkSize)
  - Work-items ID offset (offset)
  - Work groups size (localWorkSize)
  - Result Size (result)

Programming Model Extension Example

Matmul.java

```java
public class Matmul {
    public static void main(String[] args) {
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        int[][] C;
         ...
        C = multiply(A, B);
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    public static int[][] multiply (int[][] A, int[][] B) {
        // Matrix multiplication code
        //C = AB
        ...
        return C;
    }
}
```

Matmul.cl

```c
__kernel void multiply (
    __global const int *a,
    __global const int *b,
    __global int *c) {
        //Matrix multiplication code
        // C = AB
        ...
}
```

CEI.java

```java
public interface CEI {
    @OpenCL(kernel="matmul.cl", globalWorkSize="par0.x,par1.y", resultSize="par0.x,par1.y")
    @Method(declaringClass="es.bsc.compss.matmul.Matmul")
    int[][] multiply (
        @Parameter(direction = IN)
        int[][] A,
        @Parameter(direction = IN)
        int[][] B
    );
}
```

Kernel Execution Lifecycle

[0 .- Fetch input values from a remote node]

1 .- Create a cl_buffer for each parameter and the result

   cl_mem par0Dev = clCreateBuffer(…);

2 .- Fill the cl_buffers corresponding to the input values

   clEnqueueWriteBuffer(…, par0Dev, …);

3 .- Run the kernel

   cl_kernel task1 = clCreateKernel (…);

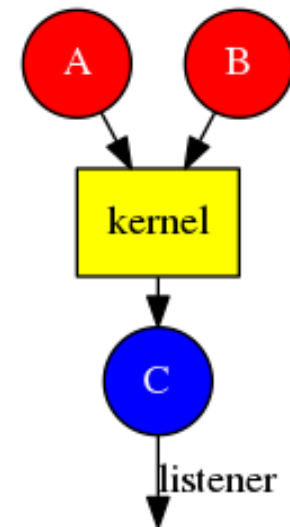   clSetKernelArgument(task1, 0, par0Dev);

   clNDRangeKernel(…, task1, …)

4 .- Bring back the modified/result values

   clEnqueueReadBuffer(…, par0Dev, …)

5 .- Notify the existence of the created values to enable data-dependent tasks executions.

Kernel Execution Lifecycle Management

- Management is twofold:
  - The runtime library directly controls the creation of OpenCL structures and Host memory management
    - Fetching data from remote workers
    - Allocating the necessary device memory space to host the buffers
    - Creating the Kernel object and setting up its arguments
    - Publishing the value existence to release data dependencies

  - It delegates on the out-of-order mode of OpenCL the execution of OpenCL commands:
    - clEnqueueWriteBuffer
    - clNDRangeKernelExecution
    - clEnqueueReadBuffer

Optimizations

- Keep values on the device memory to reuse them
- Keep track of the events generating the values

C = matmul(A,B)

E = matmul(C,D)

Executing Platform Selection

- Each task is profiled to measure the execution time and energy consumption

- Using historical values of previous executions the runtime forecasts:
  - The execution time on the resource
  - The energy consumption of the execution
  - The waiting time before the execution start

- According to a heuristic defined by the application user and based on the forecasts, the runtime picks one platform
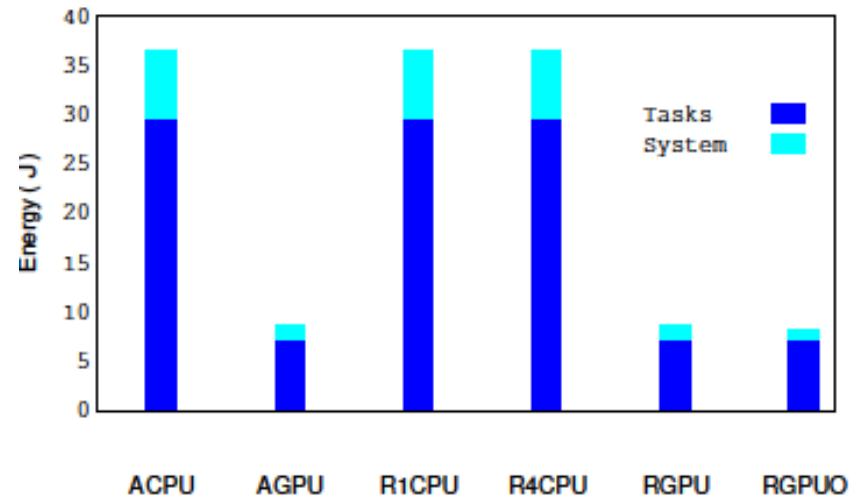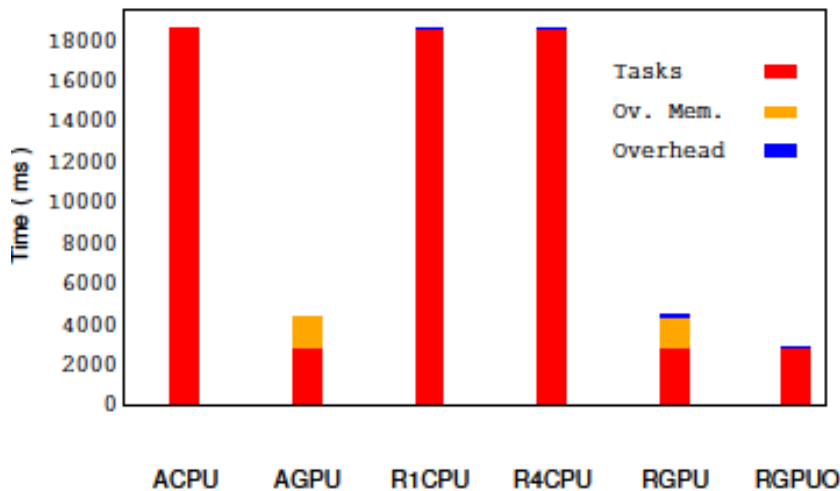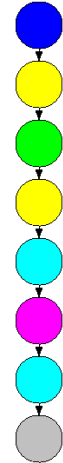
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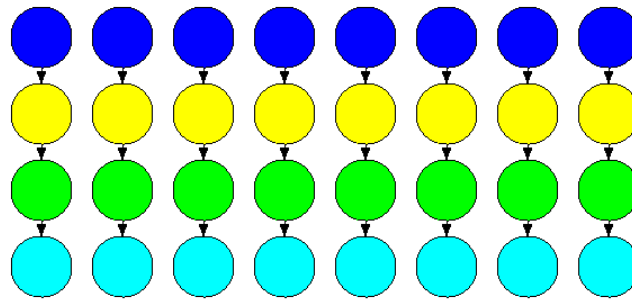**BSC**

Performance Evaluation

Digits Recognition

- CNN recognizing 512 handwritten digits

- Sequence of 8 steps (tasks) processing all 512 images

- GPU speeds up task calculation by 6.5x

- Optimizations reduce the 1,531 ms overhead to 5 ms

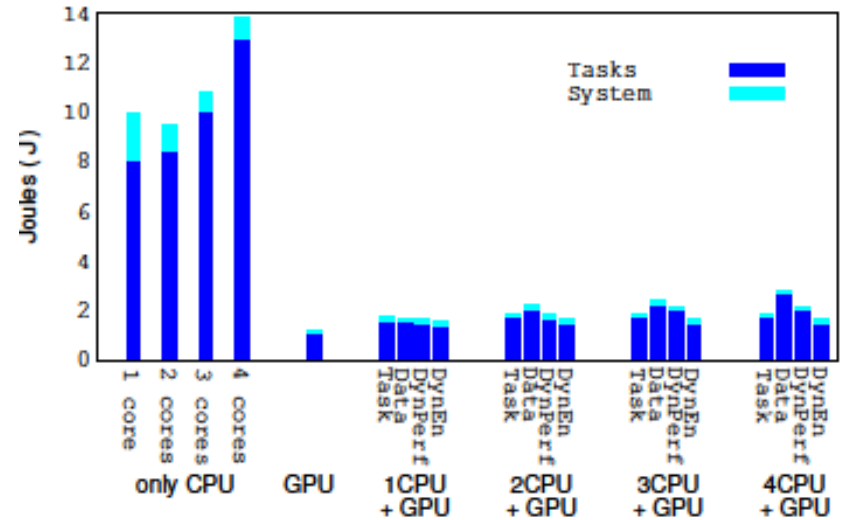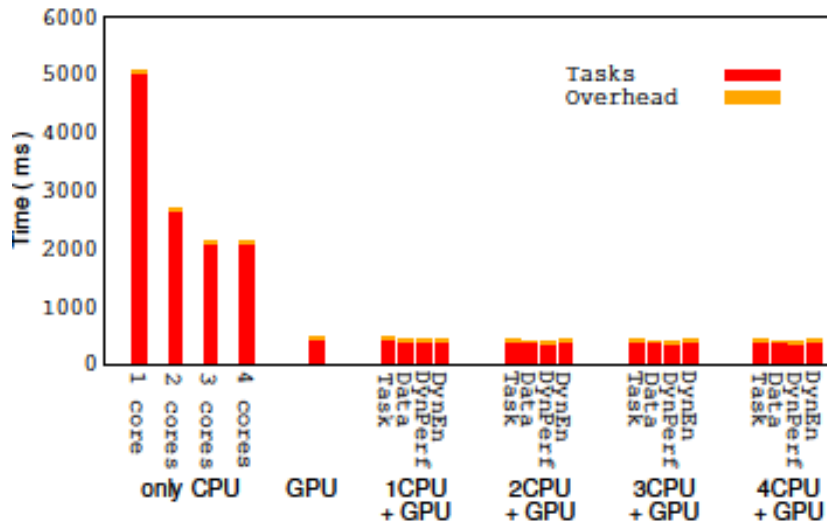- Energy consumption falls from 30 J to 8J

Canny Edge detection

- Detects the edges in 30 frames of a video
- Each frame goes through 4 stages (4 tasks)

- Developers normally try to balance the load according to:
  - Data Partitioning: assign the processing of a whole frame to a resource
  - Task Partitioning: run the two first tasks of every frame on the GPU and the last two on the CPU
- 2 dynamic policies
  - DynPerf: minimizes the execution time
  - DynEn: slows down 1 ms the execution if that saves 5 mJ

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Canny Edge detection

- GPU speeds up 12x the task computation

- GPU reduces the energy consumption from 9.89 J to 1.34 J

- Data Partitioning achieves lower execution times but consumes more energy than Task Partitioning

- DynPerf achieves the lowest execution time (375 ms)

- DynEn gets lower energy consumptions while its performance compares to Task Partitioning

Conclusions

- The programming model allows developers to code parallel applications that run on heterogeneous systems being unaware of the parallelization and collaboration issues

- Using GPUs reduces drastically the timespan and the energy consumption of the application

- Dynamic policies provide applications with
  - Portability
  - Flexibility

- Code is available at
  - http://compssdev.bsc.es/gitlab/flordan/WACCPD
  - http://compssdev.bsc.es/gitlab/flordan/COMPSs-Mobile

Future Work

- Enable the usage of the GPUs to compute tasks on remote nodes

- Improve the forecasts for the energy and end time

- Allow the dynamic policies to change an initial decision on which resources the task will run

Thank you!

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