

Benchmark porting on GPU using OpenACC

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The objective

The scope of the project is to take a program and to parallelize it using openACC.

Program: https://github.com/yuhc/gpu-rodinia/tree/master/openmp/streamcluster

Pc components:

```
Detected 1 CUDA Capable device(s)
Device 0: "GeForce 920MX"
 CUDA Driver Version / Runtime Version
                                                 9.1 / 9.1
 CUDA Capability Major/Minor version number:
                                                 5.0
 Total amount of global memory:
                                                 2048 MBytes (2147483648 bytes)
 ( 2) Multiprocessors, (128) CUDA Cores/MP:
                                                 256 CUDA Cores
 GPU Max Clock rate:
                                                 993 MHz (0.99 GHz)
 Memory Clock rate:
                                                 900 Mhz
 Memory Bus Width:
                                                 64-bit
 L2 Cache Size:
                                                 1048576 bytes
```

Produttore: ASUSTek Computer Inc.

Modello: X556UV

Processore: Intel(R) Core(TM) i7-6500U CPU @ 2.50GHz 2.59 GHz

Memoria installata (RAM): 12,0 GB (11,9 GB utilizzabile)

Note: This processor has only 2 physical cores!

Analysis of the code



Time is expressed in seconds

Analysis of the code

Analysing the time the program took to execute, it has been noticed that the most expansive time part of the program was the pgain cycle. In particular this part of the loop:

```
float x_cost = dist(points->p[i], points->p[x], points->dim)

* points->p[i].weight;

float current_cost = points->p[i].cost;

float current_cost = points->p[i].cost;

float current_cost > {

// point i would save cost just by switching to x

// (note that i cannot be a median,
// or else dist(p[i], p[x]) would be 0)

switch_membership[i] = 1;
cost_of_opening_x += x_cost - current_cost;

// cost_of_opening_x += x_cost - current assignment cost of i

// consider the savings that i's **current** median would realize
// if we reassigned that median and all its members to x;

// one we've already accounted for the fact that the median
// would save z by closing; now we have to subtract from the savings
// the extra cost of reassigning that median and its members
int assign = points->p[i].assign;
lower[center_table[assign]] += current_cost - x_cost;
}
```

Bottom-up optimization

At this point, it has been used a bottom up approach to let the GPU handle that cycle, starting from the pgain_dist loop using structured data region and kernels.

In order to manage better the memory from the host (cpu) to the device, some temporary arrays were introduced.

Because of this approach, the data were copyed in and out of the device each time they were used inside the parallel region.

This resulted in a lot of time spended in data trasfer from the CPU and the GPU.

Thus, it was decided to use unstructured data region, and «present_or_copy» clause. Since present table sometimes was giving problems, only «update» pragmas were used.

Then, data entering / exit were firstly made before pgain call (inside pFL), then inside localSearch.

Finally, compiling with -ta=tesla:managed gave reasonable results.

Bottom-up optimization Structured data region

```
memset(temp diff, 0, k2*sizeof(double));
#pragma acc data copy(temp diff[k1:k2]) copyin(dim points) copyin(points[0:1]) copyin(points->weight p[0:k2])\
copyin(points->assign p[0:k2]) copyin(points->cost p[0:k2]) copyin(points->coord p[k1:k2][0:dim points]) \
copy(switch membership[k1:k2]) copy(cost of opening x) copy(result)//copyin(current point coords[0:dim points]) copy(center table[0:centerDim])
  #pragma acc kernels// reduction(+: result)
    for (i = k1; i < k2; i++) {
      result=0.0;
          for (int j=0;j<dim points;j++)
            result += (points->coord p[i][j] - points->coord p[x][j])*(points->coord p[i][j] - points->coord p[x][j]);
      float x cost = result* points->weight p[i];
      float current cost = points->cost p[i];
      float temp = x cost - current cost;
      if ( temp < 0) {
        switch membership[i] = 1;
        cost of opening x += temp;
        temp diff[i] = temp*(-1);
```

Bottom-up optimization Unstructured data region

```
#pragma acc enter data copyin(temp_diff[k1:k2]) copyin(points[0:1])
copyin(switch membership[k1:k2]) \
pcopyin(points->weight p[0:k2]) pcopyin(points->coord p[k1:k2][0:dim points])
copyin(result) copyin(dim points) // copyin(cost of opening x)
#pragma acc declare present(points[0:1], switch membership[0:k2], temp diff[k1:k2])
#pragma acc kernels// reduction(+: result)
    for (i = k1; i < k2; i++) {
     result=0.0;
      float temp = x cost - current cost;
      if (temp < 0) {
       switch membership[i] = 1;
        cost of opening x += temp;
        temp_diff[i] = temp*(-1);
#ifdef DEBUGGER
#pragma acc exit data copyout(temp_diff[k1:k2]) copyout(switch_membership[0:k2]) \
 delete(switch membership[k1:k2]) delete(result) \
```

Bottom-up optimization Going-up: data entering before pgain

```
#pragma acc enter data copyin(points[0:1])
#pragma acc enter data copyin(points->weight p[0:k2])
#pragma acc enter data copyin(points->coord p[0:k2][0:dim points])
#pragma acc enter data create(points->assign p[0:k2], points->cost p[0:k2], switch membership[0:k2])
#ifdef DEBUGGER
fprintf(stdout, "Post acc enter data main ext loop\n");
#endif
for (i=0;i<iter;i++) {
  x = i%numfeasible:
  change += pgain(feasible[x], points, z, k, pid, barrier);
#pragma acc exit data copyout(points->weight p[0:k2])
#pragma acc exit data copyout(points->coord p[0:k2][0:dim points])
#pragma acc exit data copyout(points->assign p[0:k2], points->cost p[0:k2], switch membership[0:k2], points[0:1])
#pragma acc exit data delete(points->weight p[0:k2])
```

Bottom-up optimization Going-up: data entering before pFL

```
#pragma acc enter data copyin(points[0:1],points->weight p[0:points num],points->coord p[0:points num][0:dim points])
#pragma acc enter data copyin(temp diff[0:points num])
#pragma acc enter data create(points->assign p[0:points num], points->cost p[0:points num], switch membership[0:points num])
 while(1) {
    d++;
#ifdef PRINTINFO
    if( pid==0 )
  printf("loz = %lf, hiz = %lf\n", loz, hiz);
  printf("Running Local Search...\n");
    cost = pFL(points, feasible, numfeasible,
         z, &k, cost, (long)(ITER*kmax*log((double)kmax)), 0.1, pid, barrier);
    if (((k \le (1.1)*kmax)&&(k >= (0.9)*kmin))||
  ((k \le kmax+2)&&(k >= kmin-2))) {
#ifdef PRINTINFO
      if( pid== 0)
    printf("Trying a more accurate local search...\n");
#endif
```

Using GPU parallelism

At this point, it was tried to exploit the GPU thread/warp/blocks. The division in the bottom picture, resulted the best one after several tries.

```
#pragma acc declare present(points[0:1], switch_membership[0:k2], temp_diff[k1:k2]) \
present(points->weight_p[0:k2], points->coord_p[k1:k2][0:dim_points])

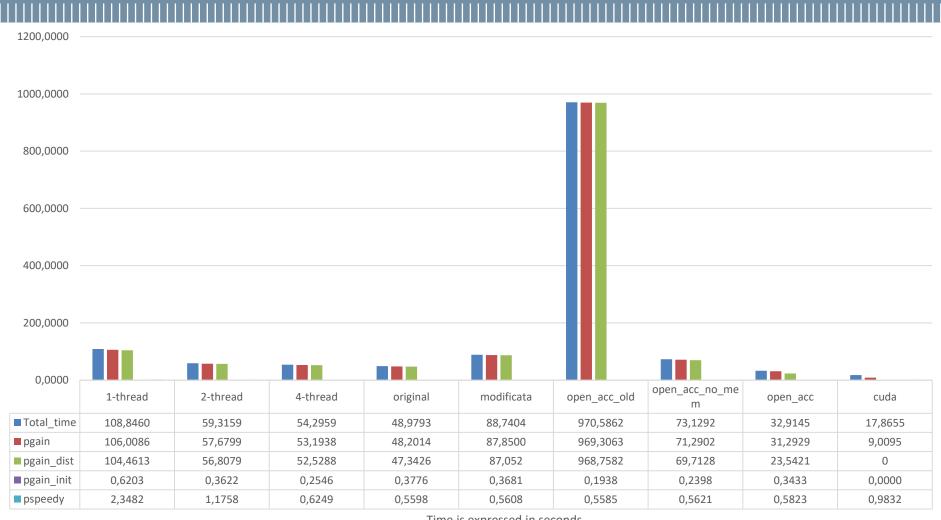
#pragma acc parallel loop reduction(+: result) device_type(nvidia) gang vector_length(64)//

for ( i = k1; i < k2; i++ ) {
    result=0.0;
    #pragma acc loop gang vector //vector_length(64)//gang(16) vector_length(32) // vector

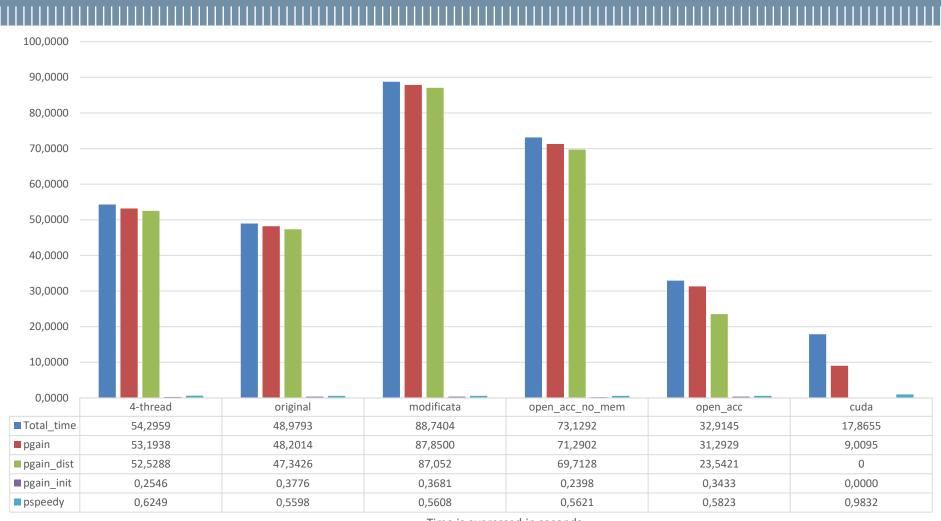
for (int j=0;j<dim_points;j++)

{
    result += (points->coord p[i][j] - points->coord p[x][j])*(points->coord p[i][j] - points->coord p[x][j]);
```

Final results



Final results - filtered



Conclusions

Since the pc where original code using threads were tested didn't have more than two physical cores, the algorithm using thread didn't perform really well. Using other CPUs, performance significantly improves arriving to near 10 seconds – which is even better that CUDA!

The limits of OpenACC is the lack of possibility to allocate exactly the amount of block/threads needed without having to do a significant refactor of the code. This is also a reason why CUDA's code works better: the code was completely rebuilt around it.

Another limit of the solution proposed, is the limit of memory usable on the GPU.

However, OpenACC focus on portability, and the version should be able to perform well on diffrent devices.