# Lab Course Efficient Programming of Multicore Processors and Supercomputers

Report 2: Sequential optimization

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#### Sequential Optimization: Matrix Access

#### Access patterns into the Matrix

- ▶ Before Optimization: column wise access in relax\_jacobi
- low spatial locality

```
1 for (j = 1; j < sizex - 1; j++) {
          for (i = 1; i < sizey - 1; i++) {
2
                   utmp[i * sizex + j] = 0.25 * (
                                   u[i * sizex + (j - 1)]
4
                                       + // left
                                   u[i * sizex + (j + 1)]
5
                                       + // right
                                   u[(i-1) * sizex + j]
                                       + // top
                                   u(i + 1) * sizex +
                                       i]); // bottom
8
9
10
11 }
```

#### Sequential Optimization: Matrix Access

Solution: change loop, row wise access

```
for (j = 1; j < sizex - 1; j++) {
          for (i = 1; i < sizey - 1; i++) {
                   utmp[i + j * sizey] = 0.25 * (
                           a[i + (j - 1)*sizey] + // left
4
5
                           a[i + (j + 1)*sizey] + //
                               right
                           a[(i - 1) + j*sizey] + // top
6
                           a[(i + 1) + j*sizey]); //
                               bottom
8
9
10
11 }
```

# Sequential Optimization: No Copy

#### **Avoid copy operations**

▶ Before Optimization: in relax\_jacobi, copy back all new values back to the array

# Sequential Optimization: No Copy

▶ Idea: Instead of recomputing, just swap the pointers

# Sequential Optimization: No Copy

Important: Need to initialize uhelp with u

▶ Note: this took us 2 days to figure out

#### Sequential Optimization: Residual

#### Calculation of Residuum

▶ Before Optimization: For residual at timestep t, compute entire relaxation of timestep t+1

```
1 for (j = 1; j < sizex - 1; j++) {
          for (i = 1; i < sizey - 1; i++) {
                   unew[i + j * sizey] = 0.25 * (
3
                           a[i + (j - 1)*sizey] + // left
                           a[i + (j + 1)*sizey] + //
5
                               right
                           a[(i - 1) + j*sizey] + // top
                           a[(i + 1) + j*sizey]); //
                               bottom
8
                           diff = unew - u[i * sizex + j];
10
                           sum += diff * diff;
12 }
```

#### Sequential Optimization: Residual

▶ Idea: Save results of previous timestep to compute residual

#### Sequential Optimization: Residual

 Second Idea: Compute relaxation directly with residual and save for later

```
1 for (i = 1; i < sizey - 1; i++) {
          for (j = 1; j < sizex - 1; j++) {
                   atmp[i * sizex + j] = 0.25 * (
3
                           a[i * sizex + (j - 1)] + //
                               left.
                           a[i * sizex + (j + 1)] + //
5
                               right
                           a[(i - 1) * sizex + j] + //
6
                               top
                           a[(i + 1) * sizex + j]); //
7
                                bottom
8
                           diff = atmp[i * sizex + j] -
9
                                a[i * sizex + j];
                            sum += diff * diff;
10
11
12
13 return sum:
```

## Sequential Optimization: Interleaving

# Cache optimization by interleaving of iterations: tiling and wavefront

- ▶ Due to the spatial dependencies between the grid points, tiling and wavefront become non-trivial
- Traditional tiling would yield wrong/old values at the borders
- Idea: Tile grid and compute one iteration on tiles
- However: Due to the already sequential access pattern, wavefront interleaving is not expected to give a significant improval of cache behaviour in a non-parallel environment

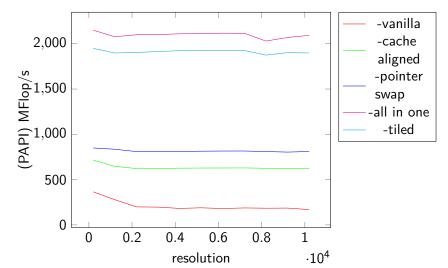
# Sequential Optimization: Interleaving

#### Tiling Solution:

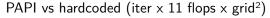
```
1 for (tiley = 1; tiley < sizey-1; tiley += tilesize){</pre>
          for (tilex = 1; tilex < sizex-1; tilex +=
               tilesize){
                   for (i = tiley; (i < sizey - 1) && (i
3
                       < tiley+tilesize); i++) {
                            for (j = tilex; (j < sizex -
4
                                1) && (j < tilex+tilesize);</pre>
                                j++) {
                                     atmp[i * sizex + j] =
5
6
```

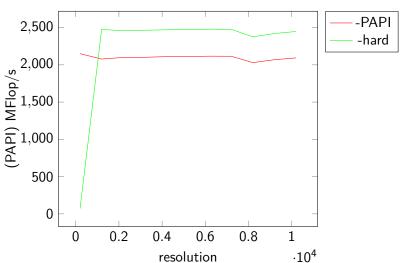
## Task 2.1 Sequential optimization - MFlop/s

Performance of Jacobi with various optimization steps (all -O2)



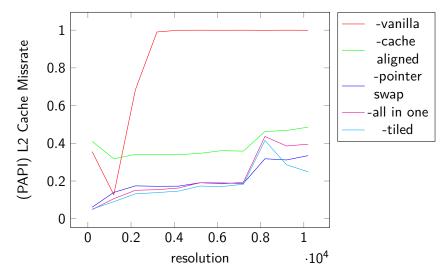
#### Task 2.1 Sequential optimization - MFlop/s





#### Task 2.1 Sequential optimization - L2 Missrate

Performance of Jacobi with various optimization steps (all -O2)



#### Task 2.2: Power of Two

There is a performance issue around sizes of powers of 2, e.g. np=1020/1022/1024/1026.

- Can you explain it?
- How to get rid of that issue?

When using array sizes of powers of two, the entries above and below will have similar least significant bits, this will cause them to be put into the same cache set. Hence, conflict misses will become more probable.

A way to get rid of this problem is array padding, so padding the arrays with a couple more entries, so that the least significant bits of neighboring cachelines differ, therefore reducing conflict misses.

Thank you for your attention!