1. Data Level Parallelism

<table>
<thead>
<tr>
<th>Function</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>__m128i_mm_load1_si128( )</td>
<td>returns 128-bit one vector</td>
</tr>
<tr>
<td>__m128i_mm_loadu_si128( __m128i *p )</td>
<td>returns 128-bit vector stored at pointer p</td>
</tr>
<tr>
<td>__m128i_mm_mul_ps(__m128 a, __m128 b)</td>
<td>returns vector (a0<em>b0, a1</em>b1, a2<em>b2, a3</em>b3)</td>
</tr>
<tr>
<td>void_mm_storeu_si128(__m128i *p, __m128i a)</td>
<td>stores 128-bit vector a at pointer p</td>
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</tbody>
</table>

1. Implement the following function, which returns the sum of two arrays:

```c
static int product_vectorized(int n, int *a) {
    int result[4];
    __m128i prod_v = __mm_load1_si128();
    for(int i = 0; i < n/4 * 4; i += 4) { // Vectorised loop
        prod_v = __mm_mul_ps(prod_v, __mm_loadu_si128((__m128i *) (a + i)));
    }

    __mm_storeu_si128((__m128i *) result, prod_v);
    for(int i = n/4 * 4; i < n; i++) { // Handle tail case
        result[0] *= a[i];
    }
    return result[0] * result[1] * result[2] * result[3];
}
```
3. **Thread Level Parallelism**

```c
#pragma omp parallelism
{ /* code here */
}
#pragma omp parallel for
for (int i = 0; i < n; i++) {
    /* code here */
}
```

1. For the following snippets of code below, circle one of the following to indicate what issue, if any, the code will experience. Then provide a short justification. Assume the default number of threads is greater than 1. Assume no thread will complete before another thread starts executing. Assume `arr` is an int array with length `n`.

   a) // Set element `i` of `arr` to `i`
   
   ```c
   #pragma omp parallel
   (int i = 0; i < n; i++)
   arr[i] = i;
   
   Sometimes incorrect  Always incorrect  Slower than serial  Faster than serial
   ```

   Slower than serial – there is no for directive, so every thread executes this loop in its entirety. `n` threads running `n` loops at the same time will actually execute in the same time as 1 thread running 1 loop. Despite the possibility of false sharing, the values should all be correct at the end of the loop. Furthermore, the existence of parallel overhead due to the extra number of threads could slow down the execution time.

   b) // Set `arr` to be an array of Fibonacci numbers.

   ```c
   arr[0] = 0;
   arr[1] = 1;
   #pragma omp parallel for
   for (int i = 2; i < n; i++)
   arr[i] = arr[i-1] + arr[i-2];
   
   Sometimes incorrect  Always incorrect  Slower than serial  Faster than serial
   ```

   Always incorrect (if `n>4`) – Loop has data dependencies, so the calculation of all threads but the first one will depend on data from the previous thread. Because we said “assume no thread will complete before another thread starts executing,” then this code will always be wrong from reading incorrect values.

   c)
// Set all elements in arr to 0;
int i;
#pragma omp parallel for
for (i = 0; i < n; i++)
    arr[i] = 0;

Sometimes incorrect Always incorrect Slower than serial Faster than serial

Faster than serial – the for directive actually automatically makes loop variables (such as the index) private, so this will work properly. The for directive splits up the iterations of the loop into continuous chunks for each thread, so no data dependencies or false sharing.

2. Consider the following code:
// Decrements element i of arr. n is a multiple of omp_get_num_threads()
#pragma omp parallel {
    int threadCount = omp_get_num_threads();
    int myThread = omp_get_thread_num();
    for (int i = 0; i < n; i++) {
        if (i % threadCount == myThread)
            arr[i] *= arr[i];
    }
}
What potential issue can arise from this code?
False sharing arises because different threads can modify elements located in the same memory block simultaneously. This is a problem because some threads may have incorrect values in their cache block when they modify the value arr[i], invalidating the cache block. A fix to this will be discussed in lab.

3. Data race and Atomic operations.

The benefits of multi-threading programming come only after you understand concurrency. Here are two most common concurrency issues:

- **Cache-incoherence**: each hardware thread has its own cache, hence data modified in one thread may not be immediately reflected in the other. The can often be solved by bypassing cache and writing directly to memory, i.e. using `volatile` keyword in many languages.
- The famous **Read-modify-write**: Read-modify-write is a very common pattern in programming. In the context of multi-thread programming, the **interleaving** of R,M,W stages often produces a lot of issues.
To solve problem with Read-modify-write, we have to rely on the idea of **undisrupted execution**.

In RISC-V, we have two categories of atomic instructions:
- Load-reserve, store-conditional (undisrupted execution across multiple instructions)
- Amo.swap (single, undisrupted memory operation) and other amo operations.
Both can be used to achieve atomic primitives, here are two examples.

### Instruction semantics:

**lr**: Loads the four bytes from memory at address x[rs1], writes them to x[rd], sign-extending the result, and registers a reservation on that memory word.

**sc**: Stores the four bytes in register x[rs2] to memory at address x[rs1], provided there exists a load reservation on that memory address. Writes 0 to x[rd] if the store succeeded, or a nonzero error code otherwise.

**Amoswap**: Atomically, let t be the value of the memory word at address x[rs1], then set that memory word to x[rs2]. Set x[rd] to the sign extension of t.

Question: why do we need special instructions for these operations? Why can’t we use normal load and store for `lr` and `sc`? Why can’t we expand amoswap to a normal load and store?

Answer: For `lr` and `sc`, after `lr`, other threads cannot write to the location marked reserve, hence the value loaded from memory (a3 in the above example) will be unchanged between `lr` and `sc`. For amoswap, it does load and store in one single CPU cycle, hence the operation is atomic and undisruptable.