

## ECE408/CS483/CSE408 Exam #1, Fall 2017

Tuesday, October 17, 2017

- You are allowed one 8.0x11.5 cheat sheet with notes on both sides. The minimal font size for your text on the cheat sheet should be 8pts.
- No interactions with humans other than course staff are allowed.
- This exam is designed to take 170 minutes to complete. To eliminate time pressure and allow for any unforeseen difficulties, we will give everyone 180 minutes.
- This exam is based on lectures, textbook chapters, as well as lab MPs/projects.
- The questions are randomly selected from the topics we covered up to and including parallel scan.
- You can write down the reasoning behind your answers for possible partial credit.
- **Good luck!**

**Question 1 (30 points, suggested time allocation 40 minutes):** multiple-choice and short-answer questions. If you get more than 30 points by answering all questions (1-9), your score will saturate at 30 points. The bonus question is extra.

For multiple-choice questions, give a concise explanation for your answer for possible partial credit. Answer each of the short-answer questions in as few words as you can. Your answer will be graded based on completeness, correctness, and conciseness.

1. (4 points) We want to use each thread to calculate four (4) output elements of a vector addition. Each block processes  $4 \times \text{blockDim.x}$  consecutive elements that form 4 sections. All threads in each block will first process a section with each thread processing one element. They will then all move to the next section with each thread processing one element. For each section, consecutive threads should process consecutive elements. What would be the kernel code expression for forming the value of  $i$ , the data index of the first element to be processed by each thread?

- (A)  $i = \text{blockIdx.x} \times \text{blockDim.x} \times 4$ ;
- (B)  $i = \text{blockIdx.x} \times \text{threadIdx.x} + \text{threadIdx.x}$
- (C)  $i = \text{blockIdx.x} \times \text{blockDim.x} \times 4 + \text{threadIdx.x}$
- (D)  $i = \text{blockIdx.x} \times \text{blockDim.x} + \text{threadIdx.x} \times 4$

Answer: (C)

Explanation: All preceding blocks cover  $(\text{blockIdx.x} \times \text{blockDim.x}) \times 4$ . The beginning element is consecutive in this case so just add  $\text{threadIdx.x}$  to it.

2. (4 points) For vector addition, assume that the vector length is 8,000, each thread calculates four (4) output elements, and the thread block size is 1,024 threads. The programmer configures the kernel launch to have a minimal number of blocks to cover all output elements. How many threads will be in the grid?

- (A) 1,024
- (B) 8,196
- (C) 2,048
- (D) 8,000

Answer: (C)

Explanation:  $\text{ceil}(8000/\text{float}(4 \times 1024)) \times 1024 = 2 \times 1024 = 2048$ . Another way to look at it is that each block covers 4096 elements. The minimal multiple of 4096 to cover 8000 is 2 (blocks). So  $2 \times 1024$  is 2048.

3. (4 points) We are to process an 565x784 (m=565 pixels in the y or vertical dimension, n=784 pixels in the x or horizontal dimension) picture with the **PictureKernel** below:

```
__global__ void PictureKernel(float* Pin, float* Pout, int m, int n) {  
  
    // Calculate the row # of the d_Pin and d_Pout element to process  
    int Row = blockIdx.y*blockDim.y + threadIdx.y;  
  
    // Calculate the column # of the d_Pin and d_Pout element to process  
    int Col = blockIdx.x*blockDim.x + threadIdx.x;  
  
    // each thread computes one element of d_Pout if in range  
    if ((Row < m) && (Col < n)) {  
        Pout[Row*n+Col] = 2*Pin[Row*n+Col];  
    }  
}
```

Assume that each block is organized as a 2D 16x32 array of threads (16 in the y dimension and 32 in the x dimension). Which of the following statements sets up the kernel configuration properly? Assume that int variable n has value 784 and int variable m has value 565. The kernel is launched with the statement `PictureKernel<<<gridDim, blockDim>>>(d_Pin, d_Pout, m, n);`

- (A) `dim3 gridDim(ceil(1, ceil(n/32)), ceil(m/16)); dim3 blockDim(1, 32 16);`
- (B) `dim3 gridDim(ceil(n/32.0), ceil(m/16.0), 1); dim3 blockDim(32, 16, 1);`
- (C) `dim3 gridDim(ceil(m/32.0), ceil(n/16.0), 1); dim3 blockDim(32, 16,1);`
- (D) `dim3 gridDim(ceil(m/16), ceil(n/32), 1); dim3 blockDim(16, 32, 1);`

**Answer: (B)**

**Explanation:** dim3 format is (x, y, z). Since n is the size of the picture in the x direction and m is the size of the y direction, we should use n to set up the x dimension and m to set up the y dimension.

4. (4 points) In Question 3, how many warps will have control divergence?

- (A) 0
- (B)  $25 \cdot 32 + 35$
- (C) 565
- (D) 576

**Answer: (C)**

Explanation: The size of the picture in the x dimension is not a multiple of 32. That means each warp in the 565 rows of the grid will have divergence. All threads in the last (bottom) 11 warps in the last (bottom) block at the right edge will all be inactive and thus no control divergence.

5. (4 points) Write some CUDA kernel code in which the threads in the same block transpose the elements of a matrix **mat** in shared memory in-place. That is, all threads should see that  $\text{mat}[i][j]$  is swapped with element  $\text{mat}[j][i]$  after the transposition. You can assume that each block has dimensions  $\text{dim3}(16,16,1)$ . Make your code as efficient in time and space as possible.

```
__shared__ float mat[16][16];
```

Answer:

```
float temp = mat[threadIdx.y][threadIdx.x];
__syncthreads();
mat[threadIdx.x][threadIdx.y] = temp;
__syncthreads();
// 2 points per (sync, swap)
```

6. (4 points) Your kernel runs on a device where each Streaming Multiprocessor has 32K registers and 64KB of shared memory. The kernel code uses 30 registers (local variables) and 5KB of shared memory. The kernel is launched with 1,920,000 threads in total organized into a square grid with dimension 50 on each side. What is the maximum number of simultaneous blocks that will run on a single SM?

Answer:

```
Each block has 768 threads, and will use 768*30 = 23,040 registers.
Each SM has 32K registers. There can only be 1 block at a time.
// calculate registers or calculate number of threads (2) correct conclusion (2)
```

7. (4 points) What are the possible values of *\*dst* after this kernel execution?

```
__global__ void race_me(char *dst) {
    dst[0] = threadIdx.x;
}

// host code
cudaMalloc(&dst, 1);
cudaMemset(dst, 3); // assign value 3 to dst
race_me<<<1,2>>>>(dst);
```

- (A) 0
- (B) 1
- (C) 0, 1
- (D) 0, 1, 3
- (E) 3

Answer: (C) There is one block with two threads whose threadIdx.x values are 0 and 1. Either value is possible since the code is not properly synchronized.

// (3) for D, (2) for explanation mentioning thread parallelism or lack of sync even with wrong answer

8. (4 points) Which of the following are design philosophies in GPU architecture?

- (1) GPUs use massive parallelism to hide stalls
- (2) GPUs have many low-latency execution units
- (3) GPUs spread the cost of managing an instruction stream across many ALUs

- (A) 1 and 2
- (B) 1 and 3
- (C) 2 and 3
- (D) all of the above

Answer: (B) GPU execution units are not designed to have low latency.

9. (4 points) Your friend has profiled a kernel and discovered that the performance is limited by the latency of the divide operator. It turns out variable d in his kernel code (below) is 1 half the time, so your friend proposed the following optimization:

```
if (1 == d) r = d; else r = 1 / d;
```

Unfortunately, the performance did not improve. In one sentence, explain why.

Answer: The code is executed by all threads and some of the threads will still succeed so now the code has control divergence.

// (3) for correct answer, (4) for control divergence

(bonus 3 points) List the errors and typos that you reported via Piazza postings, e-mails, or in person communication with Prof. Hwu.

**Question 2 (15 points, suggested time allocation 20 minutes):** CUDA Basics. For the vector addition kernel and the corresponding kernel launch code, answer each of the sub-questions below.

```
01. __global__
02. void vecAddKernel(float* A, float* B, float* C, int n) {
03.     int i = threadIdx.x + 2*blockDim.x * blockIdx.x;
04.     if(i<n) C_d[i] = A_d[i] + B_d[i];
05.     i += blockDim.x;
06.     if(i<n) C_d[i] = A_d[i] + B_d[i];
07. }

08. int vectAdd(float* A, float* B, float* C, int n) {
    // assume that size has been set to the actual length of
    // arrays A, B, and C
09.     int size = n * sizeof(float);

    10.     cudaMalloc((void **) &A_d, size);
    11.     cudaMalloc((void **) &B_d, size);
    12.     cudaMalloc((void **) &C_d, size);
    13.     cudaMemcpy(A_d, A, size, cudaMemcpyHostToDevice);
    14.     cudaMemcpy(B_d, B, size, cudaMemcpyHostToDevice);
    15.     vecAddKernel<<<ceil(n/(2*1024.0)), 1024>>>>(A_d, B_d, C_d, n);
    16.     cudaMemcpy(C, C_d, size, cudaMemcpyDeviceToHost);
    17. }
```

2(a). (2 point) Assume that the size of A, B, and C is 20,000 elements each. How many thread blocks will be generated?

Answer: 10, or correct expression

2(b). (2 point) Assume that the size of A, B, and C is 20,000 elements each. How many warps are there in each block?

Answer: 32, or correct expression

2(c) (2 point) Assume that the size of A, B, and C is 20,000 elements. How many threads will be created in the grid?

Answer: 10,240, or correct expression  
// (-1) for carrying an error from (a)

2(d) (4 points) Assume that the size of A, B, and C is 20,000 elements each. Is there any control divergence during the execution of the kernel? If so, identify the block index and warp index that causes the control divergence. Explain why or why not.

Answer: No, there is no control divergence. The last block covers 1568 elements. The first 1024 elements are covered by the first if/add statement. The 544 needs to be covered by the second if/add statement. Since 544 is multiples of 32, 17 warps will be all active and 15 warps will be in active.

// (2) for answer, (2) for explanation

// If the student answered yes, 1 point for block 9 and 1 point for warp 24 as partial credits.

2(e). (5 points) Assume that the size of A, B, and C is 40,000 elements each. Is there any control divergence during the execution of the kernel? If so, identify the line number of the statement that causes the control divergence. Explain why or why not.

Answer: No, there is no control divergence.. Since the total number of blocks is 20 and the total number of threads in the grid will be 20480, larger than the size of the arrays, Warp 9 of Block 19 will not have divergence.

// (2.5) for answer, (2.5) for explanation, (+1) for line 6

**Question 3. (15 points, suggested time allocation 25 minutes):** Your friend wants to find the sum of all elements of the **bigArr** array with GPU to boost his program performance. After he heard that you are taking ECE408, he comes to you for help, desperately. His idea is as follows: 1) load elements into shared memory of blocks; 2) use threads of each block to sum up elements in its shared memory; 3) output the sum by each block to form a new array **kernelOutput**; 4) use serial code to calculate the sum of all elements in **kernelOutput**. Assume that the kernel is launched with 1024 threads in each block.

```
#define BLOCK_SIZE = 1024
```

Global variable:     float bigArr[0..n-1] (the length is n)  
                           float kernelOutput[0..ceil(n / (2\* BLOCK\_SIZE)) - 1]

Kernel code:

```
01.  __shared__ float partialSum[2048];
02.  unsigned int bid = blockIdx.x;
03.  unsigned int t = threadIdx.x;
04.  unsigned int start = _____X_____ ;
05.  if (start + t < n)
06.      partialSum[t] = bigArr[start + t];
07.  else
08.      partialSum[t] = 0.0f;
09.  if ( _____Y_____ < n)
10.      partialSum[ _____Z_____ ] = bigArr[ _____Y_____ ];
11.  else
12.      partialSum[ _____Z_____ ] = 0.0f;
13.  for (unsigned int stride = 1; stride <= BLOCK_SIZE; stride *= 2)
14.  {
15.      __syncthreads();
16.      if (t < BLOCK_SIZE / stride)
17.          partialSum[t*stride*2] += partialSum[t*stride*2+stride];
18.  }
19.  if (t == 0)
20.      kernelOutput[bid] = partialSum[0];
```

3(a). (3 pts) Please help your friend fill codes in the blanks:

X: **2 \* bid \* BLOCK\_SIZE**

Y: **start + t + BLOCK\_SIZE**

Z: **t + BLOCK\_SIZE**

**// No partial credit**



3(b). (2 pts) Does control divergence happen in the 3rd iteration of the for-loop? Assume that the warp size is 32. (Hint: draw a picture of how the threads are mapped to the data for the next few sub-questions.)

No. All active threads are consecutive and there are multiples of 32 of them.

3(c). (3 pts) How many warps in a block are active in the 3rd iteration of the for-loop? Assume that the warp size is 32.

stride = 4. threads in active =  $1024 / 4 = 256$ . warps in active =  $256 / 32 = 8$ .

// (2) for stride, (1) for the rest

3(d). (3 pts) Do you think the method your friend uses to read the global memory is coalesced? Why?

Yes. Adjacent threads access adjacent memory.

// (2) for answer, (1) for correct defn of coalescing in explanation

3(e). (4 pts) Can we put `__syncthreads()` in the if-statement as follows? Why?

```
if (t < BLOCK_SIZE / stride) {
    __syncthreads();
    partialSum[t*stride*2] += partialSum[t*stride*2+stride];
}
```

No. Because if there exists control divergence in a warp, the threads which satisfy the if-statement will be in active and wait for the threads which do not satisfy the if-statement forever because these threads are deactivated.

(2) for No, (2) for understanding/explanation of `__syncthreads()`

**Question 4. (20 points, suggested allocation of time 35 minutes).**

Your friend Jin suggested that doing the shared matrix-matrix multiplication ( $M \times N$ ) with matrix  $M$  transposed ( $M_{\text{trans}}$ ) might increase performance in terms of memory coalesced ( $M_{\text{trans}}$  is the transposed matrix of  $M$ ). The host code would provide  $M$  in its transposed form when calling the kernel. Jin being busy could not finish the transposed shared matrix-matrix multiplication and left few blanks to fill in. His idea **may** or **may not** be correct. Note that  $M_{\text{trans}}$  and  $N$  are square matrices.

```

01  #define TILE_WIDTH 32
02
03  __global__ void sgemm(float* M_trans, float* N, float* P, int
Width) {
04
05      __shared__ float Mds[TILE_WIDTH][TILE_WIDTH];
06      __shared__ float Nds[TILE_WIDTH][TILE_WIDTH];
07
08      int bx = blockIdx.x;  int by = blockIdx.y;
09      int tx = threadIdx.x; int ty = threadIdx.y;
10
11      // Identify the row and column of the P element to work on
12      int Row = by * TILE_WIDTH + ty;
13      int Col = bx * TILE_WIDTH + tx;
14
15      float Pvalue = 0;
16      // Loop over the M and N tiles required to compute P element
17      for (int m = 0; m < (Width - 1)/TILE_WIDTH + 1; ++m) {
18
19          // Collaborative load of M and N tiles into shared memory
20          if(Row < Width && m * TILE_WIDTH + tx < Width) {
21              Mds[ty][tx] = M_trans[(m*TILE_SIZE+tx)*Width+Row];
22          } else {
23              Mds[ty][tx] = 0;
24          }
25          if(Col < Width && m * TILE_WIDTH + ty < Width) ) {
26              Nds[ty][tx] = N [(m*TILE_SIZE+ty)*Width+Col];
27          } else {
28              Nds[ty][tx] = 0;
29          }
30          __syncthreads();
31
32          if(Row < Width && Col < Width) {
33              for (int k = 0; k < TILE_WIDTH; ++k) {
34                  Pvalue += Mds[k][ty] * Nds[k][tx];

```

```

35         }
36     }
37     __syncthreads();
38 }
39 if(Row < Width && Col < Width)
40     P[Row*Width + Col] = Pvalue;
41 }

```

4(a) (6 points) Fill in the missing index calculations to make this code run correctly. If you think it is impossible to make this code to run correctly, state why. (*Hint: Draw a picture of matrix multiplication for the original layout  $M$ , study the behavior of the tiles, draw another picture of matrix multiplication based on transposed  $M_{trans}$ , adjust the use of threads in loading the tile elements to maximize global memory coalescing if necessary.*)

Answer above.

4(b) (3 points) For the **tilled matrix-matrix multiplication** ( $M \times N$ ) based on the original row-major layout, which input matrix will have coalesced access? (*Hint: Draw a picture of the tile loading access patterns.*)

- (A) M
- (B) N
- (C) Both
- (D) Neither

Answer: C

4(c) (3 points) For the **transposed ( $M$  is transposed) tiled matrix-matrix multiplication** ( $M \times N$ ) based on row-major layout, which input matrix will have coalesced access? (*Hint: Look at your picture and make sure that you adjust the tile loading index calculation to maximize coalescing if necessary.*)

- (A) M
- (B) N
- (C) Both
- (D) Neither

Answer: B

4(d) (3 points) For the **basic matrix-matrix multiplication** ( $M \times N$ ) (without tiling) based on row-major layout, which input matrix will have coalesced access? (*Hint: Draw a picture of the access patterns.*)

- a) M
- b) N
- c) Both
- d) Neither

Answer: B

4(e) (3 points) For the **transposed (M is transposed) basic matrix-matrix multiplication** ( $M \times N$ ) (without tiling) based on row-major layout, which input matrix will have coalesced access? (*Hint: Draw a picture of the access patterns.*)

- a) M
- b) N
- c) Both
- d) Neither

Answer: C

4(f) (2 points) Was Jin correct on his assumption? Suppose the transposed matrix is given. Justify your answer.

Jin is not correct. The transposition access used above leads to uncoalesced accesses.

**Question 5. (20 points, suggested time allocation 40 minutes) Convolution Fun.** After teaching your roommate how to do parallel convolution, they come up with the following kernel. Their kernel is similar to the Strategy 2 presented in class where enough threads are instantiated to load all the input elements of a tile in one round. There are three major differences. Each thread block calculates two adjacent tiles in the x-dimension and the tiles and mask are no longer cubes. Look at the code below and answer the following questions.

```
#define TILE_X 8                // Output tile width in the X-dimension
#define TILE_Y 4                // Output tile width in the Y-dimension
#define TILE_Z 2                // Output tile width in the Z-dimension
#define NUM_X 2
#define X_MASK_WIDTH 3
#define Y_MASK_WIDTH 3
#define Z_MASK_WIDTH 5
#define MASK_SIZE X_MASK_WIDTH * Y_MASK_WIDTH * Z_MASK_WIDTH

__constant__ float mask[Z_MASK_WIDTH][Y_MASK_WIDTH][X_MASK_WIDTH];

__global__ void conv3d(float *input, float *output, const int z_size, const int y_size, const int
x_size) {
    __shared__ float inputTile
[TILE_Z+Z_MASK_WIDTH-1][TILE_Y+Y_MASK_WIDTH-1][TILE_X+X_MASK_WIDTH-1];
    int tx = threadIdx.x; int ty = threadIdx.y; int tz = threadIdx.z;
    int bx = blockIdx.x; int by = blockIdx.y; int bz = blockIdx.z;

    for(int x_tile = 0; x_tile < NUM_X; x_tile++) {
        (2) __syncthreads(); (solution for part B)
        int x_o = NUM_X * bx * TILE_X + tx + x_tile * TILE_X;
        int y_o = by * TILE_Y + ty;
        int z_o = bz * TILE_Z + tz;
        (2) __syncthreads(); (solution for part B)
        int x_i = x_o - X_MASK_WIDTH/2;
        int y_i = y_o - Y_MASK_WIDTH/2;
        int z_i = z_o - Z_MASK_WIDTH/2;
        (2) __syncthreads(); (solution for part B)
        if (x_i >= 0 && y_i >= 0 && z_i >= 0 && x_i < x_size && y_i < y_size && z_i < z_size)
            inputTile[tz][ty][tx] = input[(z_i * y_size + y_i) * x_size + x_i];
        else
            inputTile[tz][ty][tx] = 0.0;
        (1) __syncthreads(); (solution for part B)
        float acc = 0.0;
        (1) __syncthreads(); (solution for part B)
```

```

if(tz < TILE_Z && ty < TILE_Y && tx < TILE_X) {

    for(int z_mask = 0; z_mask < Z_MASK_WIDTH; z_mask++) {

        for(int y_mask = 0; y_mask < Y_MASK_WIDTH; y_mask++) {

            for(int x_mask = 0; x_mask < X_MASK_WIDTH; x_mask++) {

                acc += maskI[z_mask][y_mask][x_mask] * inputTile[tz+z_mask][ty+y_mask][tx+x_mask];

            }

        }

    }

    if(z_o < z_size && y_o < y_size && x_o < x_size)
        output[(z_o * y_size + y_o) * x_size + x_o] = acc;

}
(2) __syncthreads(); (solution for part B)
}

}

```

5(a) (4 points) Write out the host code for declaring the grid and block dimensions below.

Answer:

```

dim3 DimGrid((x_size/TILE_X*2.0), y_size/TILE_Y, z_size/TILE_Z);
if(x_size%(TILE_X*2)) DimGrid.x++;
if(y_size%TILE_Y) DimGrid.y++;
if(z_size%TILE_Z) DimGrid.z++;
dim3 DimBlock(TILE_X+X_MASK_WIDTH-1, TILE_Y+Y_MASK_WIDTH-1,
TILE_Z+Z_MASK_WIDTH-1);

```

5(b) (4 points) Unfortunately your roommate wasn't paying attention when you talked about \_\_syncthreads() and when you should use it. Please insert only the necessary \_\_syncthreads() to the kernel above to make sure it works.

Answer: Need 2 \_\_syncthreads(), 1 (1) and 1 (2). +2 For each.

5(c) (3 points) For an internal output tile, what is the average number of times that each input element will be accessed from the shared memory during the calculation an output tile? (Note:

*We are referring to tiles in this case, not thread blocks. Each thread block calculates two adjacent output tiles.)*

Answer:

$$(\text{Num Output elements}) * (\text{Size of mask}) / (\text{Num. Input Elements}) = \\ (8*4*2)*(3*3*5) / ((8+3-1)*(4+3-1)*(2+5-1)) = 8$$

5(d) (3 points) How does your answer to (C) compare to an output tile that is a cube that contains the same volume and why? Note: We are referring to tiles in this case, not thread blocks. Each thread block calculates two adjacent output tiles. *(Hint: Are all the input tile elements reused the same number of times? Does it matter where they are in the input tile?)*

Answer:

$$\text{A cube tile would have reuse of } (4^3)*(3*3*5) / ((4+3-1)*(4+3-1)*(4+5-1)) = 10$$

A rectangular prism has a smaller average than a cube with the same volume because there are more elements in the center that use more input elements in a cube than a rectangular prism. That is, a rectangular prism has more surface elements than a cube of the same volume. As a result, the volume of the input tile for a cube is larger than the volume of the input tile for a rectangular prism. More input elements need to be loaded to calculate the same number of output elements.

+3 For correct explanation (calculation does not need to be included)

+3 For correct reuse calculation and conclusion that cube has more reuse

+2 For reuse calculation with mistakes and conclusion that cube is better

+1 For correct reuse calculation, but incorrect conclusion

+0 For incorrect conclusion

5(e) (3 points) Consider another output tile with dimensions 2x4x8. Is there a difference in terms of execution efficiency of the kernel between this 2x4x8 output tile and the output tile in the kernel above with dimensions 8x4x2? If so explain why, if not, explain why not.

Answer: Because the elements are stored along the x then y dimensions, the 2x4x8 tile will have better memory coalescing than the 8x4x2. (As long as the assumption for which dimension is x is clear than either 2x4x8 or 8x4x2 would be accepted)

+3 Talks about what dimension needs to be larger and memory coalescing

+2 For x dimension being larger but no memory coalescing

+1 For memory reuse, as it does have an impact, but when comparing DRAM accesses, the one with lower reuse, i.e. 8 in the x, 4 in the y, 2 in the z will only access DRAM  $6 * 6$  times for each tile while the one with higher reuse, 2 in the x, 4 in the y, and 8 in the z will access DRAM  $6 * 12$  times for each tile which is twice as many accesses.

+1 For talking about control divergence as it again ignores memory coalescing. The one with 8 in the x, 4 in the y, 2 in the z will have less control divergence, but more warps.

+0 For all other answers.

5(f) (3 points) Is there any inefficiency in loading the input tiles for the two adjacent output tiles processed by each block? If so, identify the cause of the inefficiency and briefly suggest a strategy to eliminate the inefficiency. (*Hint: Draw a picture of the input tiles for these two output tiles.*)

Answer:

The left halo surface elements are loaded redundantly during the second iteration of the outermost loop. One should use a larger output tile rather than iterating over two output tiles. Or one could try to use if-statements to avoid the redundant loads during the second iteration. .

+3 A student will receive full credit if they spot this inefficiency and choose one of these strategies.

+2 For an inefficiency that is particular to the loading of input tiles for the two adjacent output tiles and no correct solution.

+1 For an inefficiency that is not particular to loading two adjacent output tiles, i.e. it is also present in the 1 tile per block case and correct solution.

+1 For a reasonable inefficiency, but no correct or incorrect solution.

+0 for an incorrect inefficiency.