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## CUDA并行计算基础

- · GPU的存储单元
- 矩阵相乘
- 矩阵转置

## GPU的存储单元

#### Each thread can:

R/W per-thread registers

R/W per-thread local memory

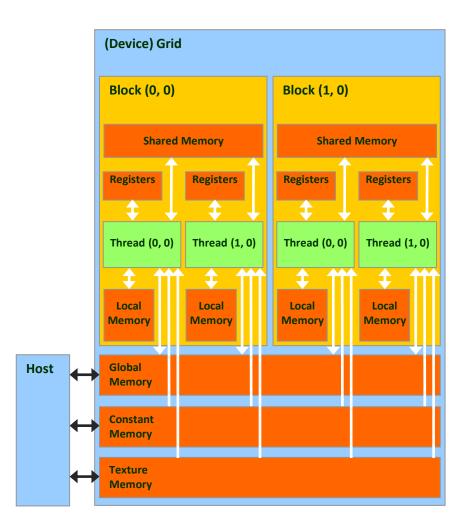
R/W per-block shared memory

R/W per-grid global memory

Read only per-grid constant memory

Read only per-grid texture memory

The host global constant texture



#### MEMORY ALLOCATION / RELEASE

CPU memory:

GPU memory:

malloc()

– cudaMalloc()

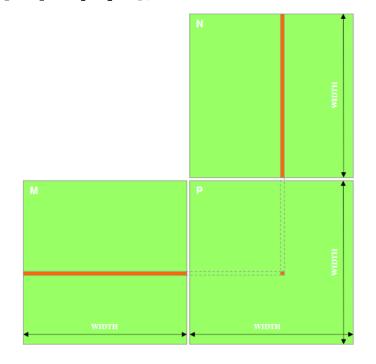
— memset()

— cudaMemset()

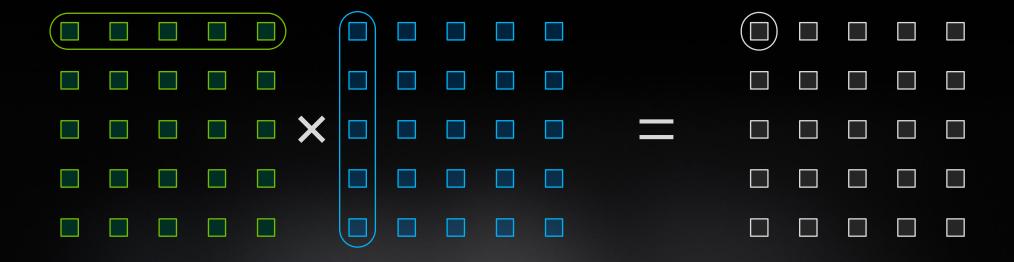
- free()

- cudaFree()

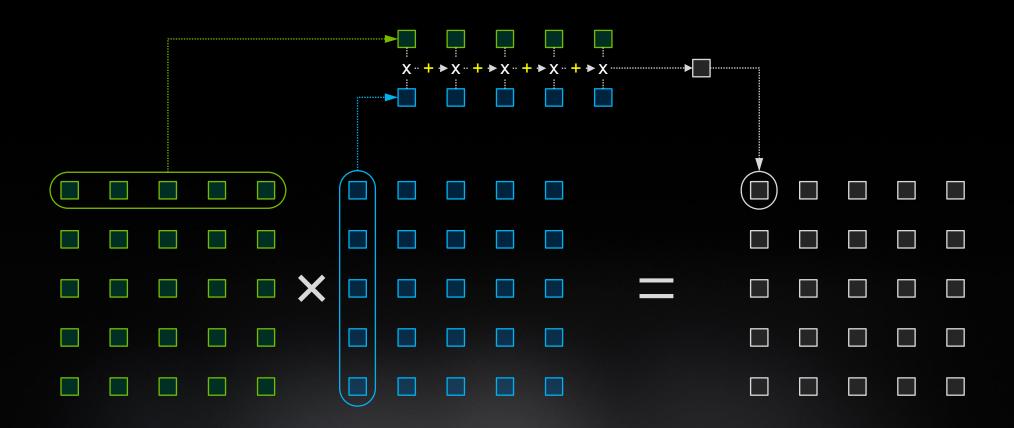
```
void cpu_matrix_mult(int *h_m, int *h_n, int *h_result, int m,
int n, int k) {
  for (int i = 0; i < m; ++i)
    for (int j = 0; j < k; ++j)
      int tmp = 0.0;
      for (int h = 0; h < n; ++h)
        tmp += h_m[i * n + h] * h_n[h * k + j];
      h result[i * k + j] = tmp;
                                P = M * N
                                假定 M and N 是方阵
```

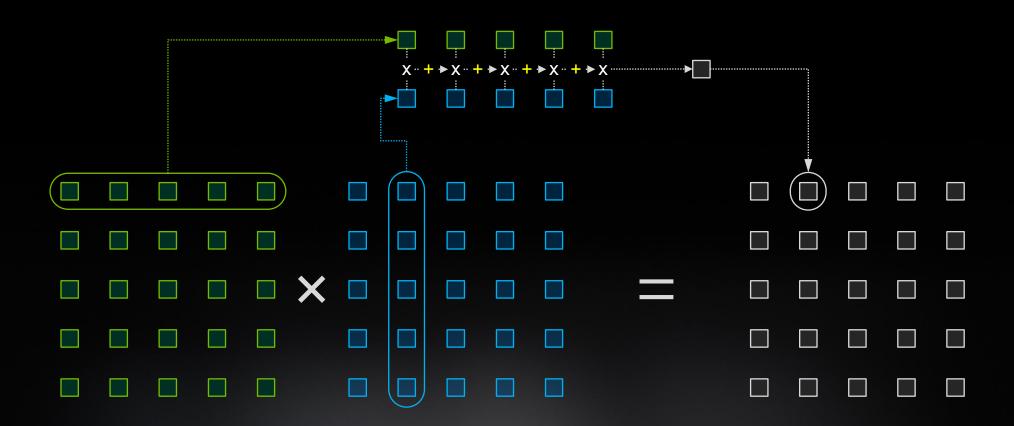


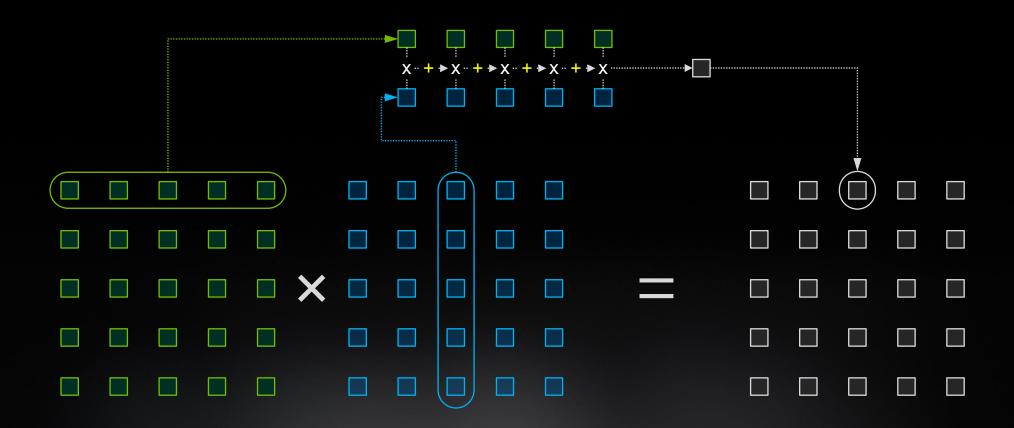
$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \times \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{bmatrix} = \begin{bmatrix} a_{11} \times b_{11} + a_{12} \times b_{21} + a_{13} \times b_{31} & a_{11} \times b_{12} + a_{12} \times b_{22} + a_{13} \times b_{32} & a_{11} \times b_{13} + a_{12} \times b_{23} + a_{13} \times b_{33} \\ a_{21} \times b_{11} + a_{22} \times b_{21} + a_{23} \times b_{31} & a_{21} \times b_{12} + a_{22} \times b_{22} + a_{23} \times b_{32} & a_{21} \times b_{13} + a_{22} \times b_{23} + a_{23} \times b_{33} \\ a_{31} \times b_{11} + a_{32} \times b_{21} + a_{33} \times b_{31} & a_{31} \times b_{12} + a_{32} \times b_{22} + a_{33} \times b_{32} & a_{31} \times b_{13} + a_{32} \times b_{23} + a_{33} \times b_{33} \end{bmatrix}$$

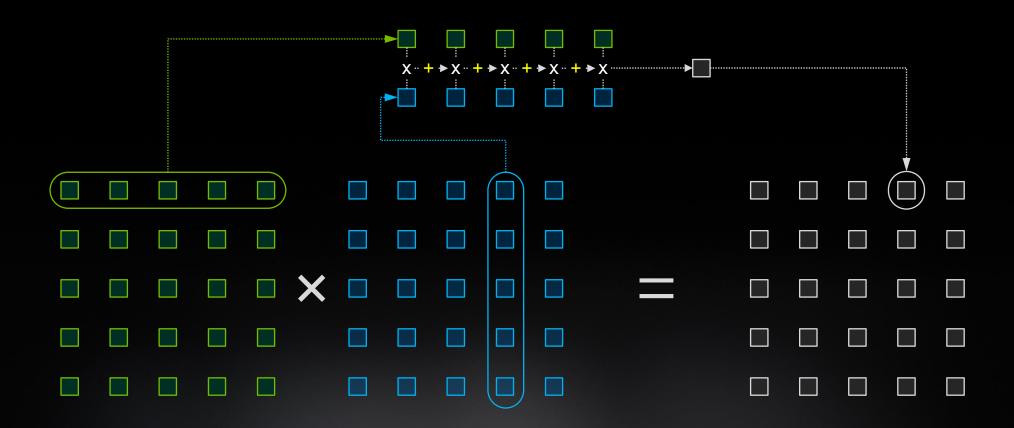


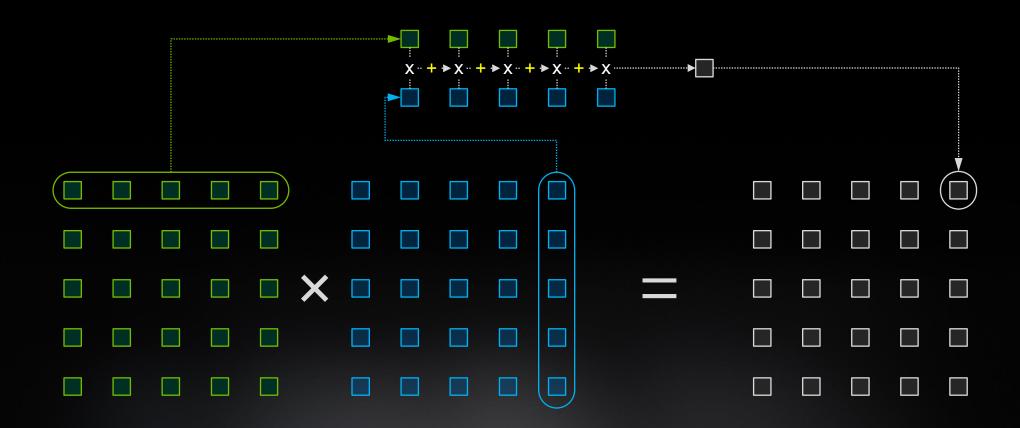








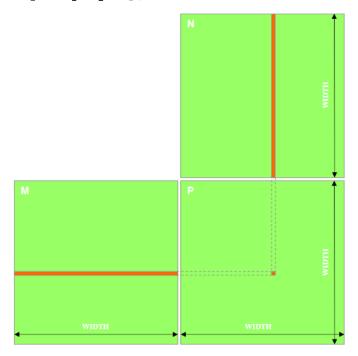




```
void cpu_matrix_mult(int *h_m, int *h_n, int *h_result, int m,
                                                                                                                      Loop 1: j: 0
int n, int k) {
  for (int i = 0; i < m; ++i)
                                                                                                                      Loop 2: i: 0
                                                                                                                      Loop 3: k: 0
    for (int j = 0; j < k; ++j)
      int tmp = 0.0;
                                                                                                +=
      for (int h = 0; h < n; ++h)
        tmp += h_m[i * n + h] * h_n[h * k + j];
      h result[i * k + j] = tmp;
                                P = M * N
                                假定 M and N 是方阵
```

如果, M=N=1000 那么, 我们要做1000\*1000\*1000次: tmp += h\_m[i \* n + h] \* h\_n[h \* k + j];

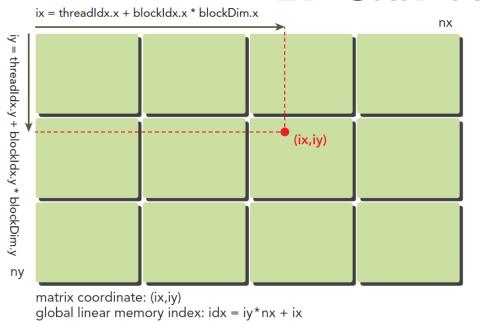
```
void cpu_matrix_mult(int *h_m, int *h_n, int *h_result, int m,
int n, int k) {
  for (int i = 0; i < m; ++i)
    for (int j = 0; j < k; ++j)
      int tmp = 0.0;
      for (int h = 0; h < n; ++h)
        tmp += h_m[i * n + h] * h_n[h * k + j];
      h result[i * k + j] = tmp;
                                P = M * N
                                假定 M and N 是方阵
```



那么,利用CUDA该怎么解决这个问题呢? 空间换时间!

#### **2D GRID AND 2D BLOCKS**

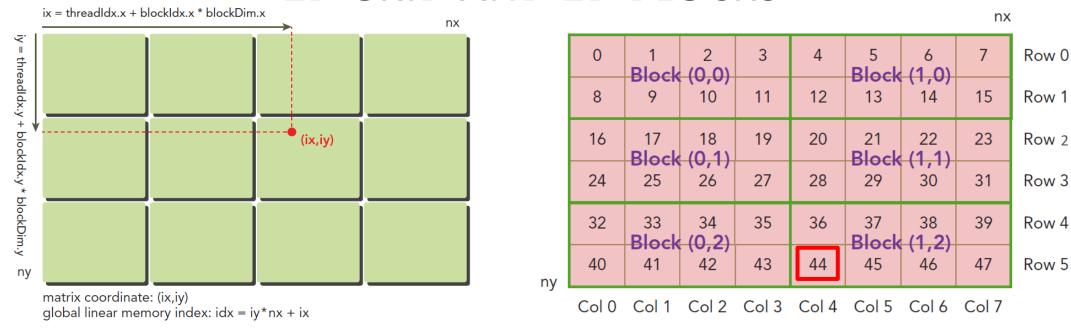
ny



0	1 Block	2 ( <b>0,0</b> )	3	4	5 Block	6 (1,0)	7	Row 0
8	9	10	11	12	13	14	15	Row 1
16	17 Block	18 ( <b>0,1</b> )	19	20	21 Block	22 (1,1)	23	Row 2
24	25	26	27	28	29	30	31	Row 3
32	33 Block	34 ( <b>0,2</b> )	35	36	37 Block	38 ( <b>1,2</b> )	39	Row 4
40	41	42	43	44	45	46	47	Row 5
Col 0	Col 1	Col 2	Col 3	Col 4	Col 5	Col 6	Col 7	•

nx

#### **2D GRID AND 2D BLOCKS**



int index = (blockIdx.y\*blockDim.y+threadIdx.y)\*nx+ (blockIdx.x \* blockDim.x + threadIdx.x);

Index == 
$$(2*2+1)*8+(1*4+0)$$

Grid												
0,0	1,0	2,0	3,0	0,0	1,0	2,0	3,0	0,0	1,0	2,0	3,0	
0,1	1,1	2,1	3,1	0,0	1,1	2,1	3,1	0,1	1,1	2,1	3,1	
0,2	1,2	2,2	3,2	0,2	1,2	2,2	3,2	0,2	1,2	2,2	3,2	
0,0	1,1	2,0	3,0	0,0	1,0	2,0	3,0	0,0	1,0	2,0	3,0	
0,0	1,1	2,1	3,1	0,0	1,1	2,1	3,1	0,1	1,1	2,1	3,1	
0,0	1,2	2,2	3,2	0,0	1,2	2,2	3,2	0,2	1,2	2,2	3,2	

每个颜色对应一个block 该线程在grid中所有线程的索引为:

Thread\_x = blockldx.x\*blockDim.x+threadIdx.x = 6
Thread\_y = blockldx.y\*blockDim.y+threadIdx.y = 3

threadIdx.x = 2 threadIdx.y = 0 blockIdx.x = 1 blockIdx.y = 1 blockDim.x = 4 blockDim.y = 3

#### 当把整个grid对应到一个矩阵的时候,它就像下面的样子:

0,0	1,0	2,0	3,0	4,0	5,0	6,0	7,0	8,0	9,0	10,0	11,0
0,1	1,1	2,1	3,1	4,1	5,1	6,1	7,1	8,1	9,1	10,1	11,1
0,2	1,2	2,2	3,2	4,2	5,2	6,2	7,2	8,2	9,2	10,2	11,2
0,3	1,3	2,3	3,3	4,3	5,3	6,3	7,3	8,3	9,3	10,3	11,3
0,4	1,4	2,4	3,4	4,4	5,4	6,4	7,4	8,4	9,4	10,4	11,4
0,5	1,5	2,5	3,5	4,5	5,5	6,5	7,5	8,5	9,5	10,5	11,5

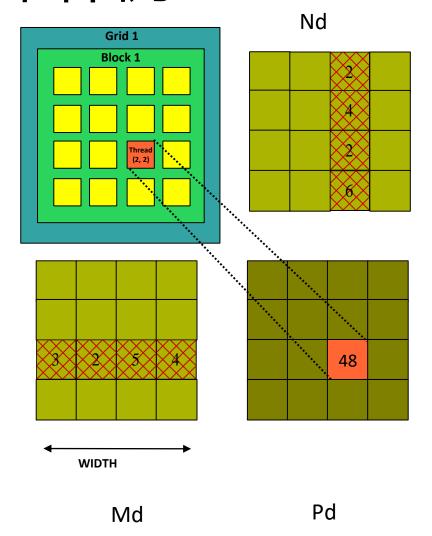
而该线程在grid中所有线程的索引正好对应到矩阵的坐标。 该线程在grid中所有线程的索引为:

Thread\_x = blockIdx.x\*blockDim.x+threadIdx.x = 6

Thread\_y = blockIdx.y\*blockDim.y+threadIdx.y = 3



- 一个线程grid计算Pd
  - □ 每个线程计算Pd的一个元素
- 每个线程
  - □读入矩阵Md的一行
  - □读入矩阵Nd的一列
  - □ 为每对Md和Nd元素执行一 次乘法和加法



```
global void gpu matrix mult(int *M,int *N, int *P, int
                                    m size, int n size, int k size)
                                      nt row = blockldx.y * blockDim.y + threadldx.y;
计算出当前执行的线程在所有
线程中的坐标
                                      nt col = blockIdx.x * blockDim.x + threadIdx.x;
                                      int sum = 0;
                                      f( col < k_size && row < m_size)
                                        for(int i = 0; i < n; i++)
                                          sum += M[row * n_size + i] * N[i * k_size + col];
读取M矩阵的一行,N矩阵的
一列,并做乘积累加
                                        P[row * k_size + col] = sum;
```

```
int main(int argc, char const *argv[])
   int m=100:
   int n=100;
   int k=100;
   int *h_a, *h_b, *h_c, *h_cc;
   cudaMallocHost((void **) &h a, sizeof(int)*m*n);
   cudaMallocHost((void **) &h_b, sizeof(int)*n*k);
   cudaMallocHost((void **) &h c, sizeof(int)*m*k);
   cudaMallocHost((void **) &h cc, sizeof(int)*m*k);
   for (int i = 0; i < m; ++i) {
       for (int j = 0; j < n; ++j) {
           h = [i * n + j] = rand() % 1024;
   }
   for (int i = 0; i < n; ++i) {
       for (int i = 0; i < k; ++i) {
           h b[i * k + j] = rand() % 1024;
   }
   int *d_a, *d_b, *d_c;
   cudaMalloc((void **) &d a, sizeof(int)*m*n);
   cudaMalloc((void **) &d b, sizeof(int)*n*k);
   cudaMalloc((void **) &d c, sizeof(int)*m*k);
   // copy matrix A and B from host to device memory
   cudaMemcpy(d a, h a, sizeof(int)*m*n, cudaMemcpyHostToDevice);
    cudaMemcpy(d b, h b, sizeof(int)*n*k, cudaMemcpyHostToDevice);
    unsigned int grid_rows = (m + BLOCK_SIZE - 1) / BLOCK_SIZE;
    unsigned int grid cols = (k + BLOCK SIZE - 1) / BLOCK SIZE;
    dim3 dimGrid(grid_cols, grid_rows);
    dim3 dimBlock(BLOCK SIZE, BLOCK SIZE);
   gpu_matrix_mult<<<dimGrid, dimBlock>>>(d_a, d_b, d_c, m, n, k);
    cudaMemcpy(h_c, d_c, sizeof(int)*m*k, cudaMemcpyDeviceToHost);
   //cudaThreadSynchronize();
```

```
__global__ void gpu_matrix_mult(int *a,int *b, int *c, int m, int n, int k)
{
    int row = blockIdx.y * blockDim.y + threadIdx.y;
    int col = blockIdx.x * blockDim.x + threadIdx.x;
    int sum = 0;
    if( col < k && row < m)
    {
        for(int i = 0; i < n; i++)
        {
            sum += a[row * n + i] * b[i * k + col];
        }
        c[row * k + col] = sum;
    }
}</pre>
```

- 在算法实现中最主要的性能问题是什么?
- 主要的限制是什么?

# 矩阵转置示例

	blo	ockD	im.x	* blc	ckld	x.x		thre	eadIdx.x					
0, ~	0, 1	0, 2	0, 3	0, 4	0, 5	0, 6	0, 7	0, 8	0, 9 0, 10	), 11	0, 12	0, 13	0, 14	0, 15
1, 0	1, :	1, 2	1, 3	1, 4	1, 5	1, 6	1, 7	1, 8	1, 9 1, 10	L, 11	1, 12	1, 13	1, 14	1, 15
2, 0	2, 1	2, 3	2, 3	2, 4	2, 5	2, 6	2, 7	2, 8	2, 9 2, 10 2	2, 11	2, 12	2, 13	2, 14	2, 15
3, 0	3, 1	3, 2	3, 3	3, 4	3, 5	3, 6	3, 7	3, 8	3, 9 3, 10	3, 11	3, 12	3, 13	3, 14	3, 15
4, 0	4, 1	4, 2	4, 3	4,	4, 5	4, 6	4, 7	4, 8	4, 9 4, 10	1, 11	4, 12	4, 13	4, 14	4, 15
5, 0	5, 1	5, 2	5, 3	5, 4	5, 5	5, 6	5, 7	5, 8	5, 9 5, 10	5, 11	5, 12	5, 13	5, 14	5, 15
6, 0	6, 1	6, 2	6, 3	6, 4	6, 5	6, ^	6, 7	6, 8	6, 9 6, 10	6, 11	6, 12	6, 13	6, 14	6, 15
7, 0	7, 1	7, 2	7, 3	7, 4	7, 5	7, 6	7, 7	7, 8	7, 9 7, 10	7, 11	7, 12	7, 13	7, 14	7, 15
8, 0	8, 1	8, 2	8, 3	8, 4	8, 5	8, 6	8, 7	8,	8, 9 8, 10	3, 11	8, 12	8, 13	8, 14	8, 15
9, 0	9, 1	9, 2	9, 3	9, 4	9, 5	9, 6	9, 7	9, 8	9, 9, 10	9, 11	9, 12	9, 13	9, 14	9, 15
10, 0	10, 1	10, 2	10, 3	10, 4	10, 5	10, 6	10, 7	10, 8	10, 9 10, 21	.0, 11	10, 12	10, 13	10, 14	10, 15
11, 0	11, 1	11, 2	11, 3	11, 4	11, 5	11, 6	11, 7	11, 8	11, 9 11, 10	.1, _1	11, 12	11, 13	11, 14	11, 15
12, 0	12, 1	12, 2	12, 3	12, 4	12, 5	12, 6	12, 7	12, 8	12, 9 12, 10 1	.2, 11	12, _3	12, 13	12, 14	12, 15
13, 0	13, 1	13, 2	13, 3	13, 4	13, 5	13, 6	13, 7	13, 8	13, 9 13, 10 1	.3, 11	13, 12	13, _3	13, 14	13, 15
14, 0	14, 1	14, 2	14, 3	14, 4	14, 5	14, 6	14, 7	14, 8	14, 9 14, 10 1	.4, 11	14, 12	14, 13	14,	14, 15
15, 0	15, 1	15, 2	15, 3	15, 4	15, 5	15, 6	15, 7	15, 8	15, 9 15, 10 1	.5, 11	15, 12	15, 13	15, 14	15, _5

```
__global__ void gpu_transpose(int *in,int *out, int width)
{
    int y = blockIdx.y * blockDim.y + threadIdx.y;
    int x = blockIdx.x * blockDim.x + threadIdx.x;

    if( y < width && x < width)
    {
        out[x * width + y] = in[y * width + x];
    }
}</pre>
```

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