#### Flatten 2D matrix

2D matrix to 1D array and back again

C++ uses **row major order**: n x m, which are the number of rows and columns also called the height and the width

To get back to 2D matrix from A(k)

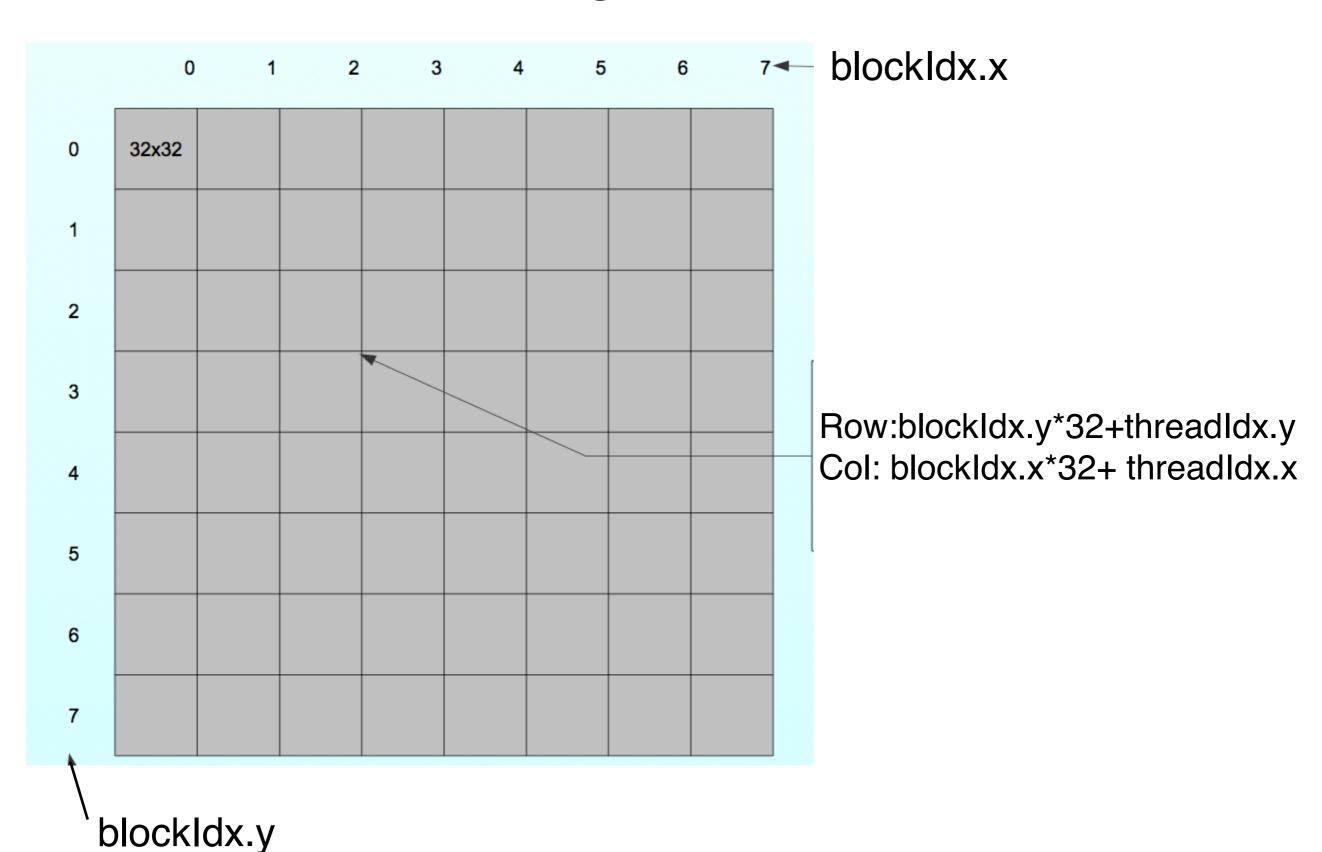
```
i=k/m; //rounding down j=k-(i^*m); or j=k\%m (where modulus gives remainder)
```

#### Matrix Copy

Problem: copy matrix a(n,m) into b(n,m). Here n=m=256; multiple of 32

```
Solution: matcopy.cu with flattened matrices
  _global___ void copymat(float * input, float * output)
int x = blockldx.x * blockDim.x + threadldx.x;
                                                 //using 2-D location in matrix
int y = blockldx.y * blockDim.y + threadIdx.y;
int length = gridDim.x*blockDim.x; //width of a row
output[y*length+x] = input[y*length+x];
int main(){
 dim3 block(32,32);
                            //NOTE: can not use block(32,32,0)
 dim3 gridDim(8,8); //8 \times 32 = 256 (perfect fit)
 copymat<<<gri>dDim,block>>>(d_input, d_output);
```

# Matrix: gridDim(8,8)



#### Matrix Copy

Instead of an 8x8 grid of 32 x 32 blocks, use 32x8 blocks four times in y direction; grid stride.

```
dim3 block(32,32);
dim3 gridDim(2,8);
copymat<<<gridDim,block>>>(d_input, d_output);
```

What is the kernel?

Why do it? Thread reuse—it is actually faster.

# Matrix Copy by 4 (using grid stride in y)

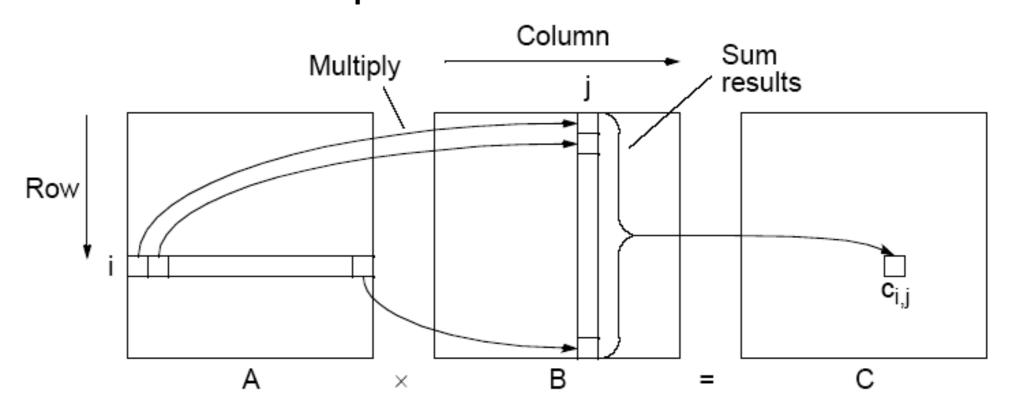
```
__global__ void copymat(float * input, float * output)
{
  int x = blockIdx.x * blockDim.x + threadIdx.x;
  int y = blockIdx.y * blockDim.y + threadIdx.y;

int length = gridDim.x*blockDim.x;

for (int j=0; j < 4*gridDim.y*blockDim.y; j+=gridDim.y*blockDim.y)
    output[(y+j)*length+x] = input[(y+j)*length+x];
}</pre>
```

#### Matrix Multiply

#### Two square matrices: N x N



$$c_{i,j} = \sum_{k=0}^{N-1} a_{i,k} * b_{k,j}$$

# Square Matrix Multiply

Simple matrix multiply with square matrices: C=A\*B with size WIDTH\*WIDTH

Procedure: row y of A times column x of B = C element (y,x)В Note that A rows are read WIDTH times; same with cols. C A.width B.width

#### C++ Code

```
for (i = 0; i < N; i++) 

for (j = 0; j < N; j++) { 

c[i][j] = 0; 

for (k = 0; k < N; k++) 

c[i][j] = c[i][j] + a[i][k] * b[k][j]; }
```

Requires n<sup>3</sup> multiplications and n<sup>3</sup> additions

# CUDA with flattened a, b, and c one thread for each c<sub>i,i</sub>

#define WIDTH 256

```
global___ void mult_mat(float * a, float * b, float * c)
int x = blockldx.x * blockDim.x + threadldx.x; //location in c
int y = blockldx.y * blockDim.y + threadldx.y;
int idx = y*WIDTH + x;
if (idx <WIDTH*WIDTH)
  float Cvalue=0.0;
  for (int i=0; i<WIDTH; i++)
     Cvalue += a[y*WIDTH + i] * b[i*WIDTH+x];
   c[y*WIDTH+x]=Cvalue;
```

#### Using Shared Memory

Option: Put all of matrix a and b into shared memory

Problem: only one block possible as memory only shared in block

Option: Put all of a row into a shared memory; better than before

Problem: reuse of all the columns

Option: Use tiles.

# Tiled Matrix Multiply

Use blocks:

load first A block and the first B block into shared memory; multiply; unload shared memory

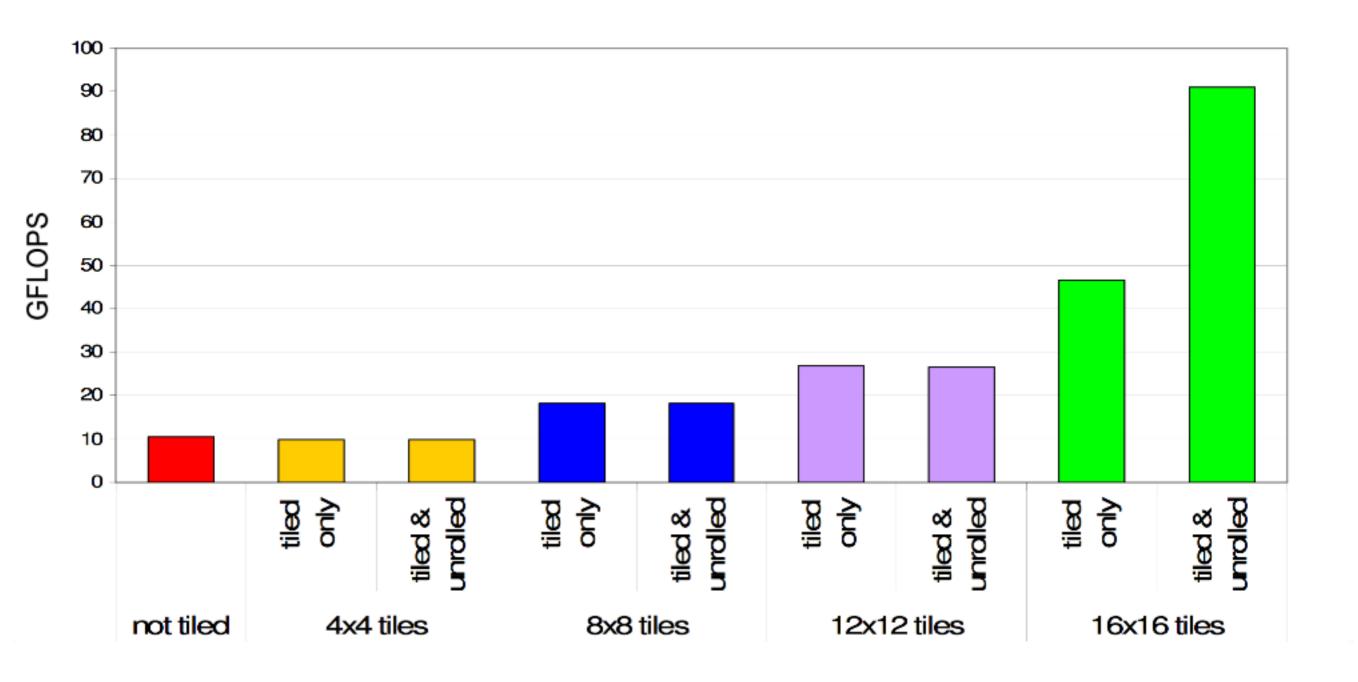
load second A block and the second B block; multiply; unload

blockCol add sums В C BLOCK\_SIZE-1 BLOCK\_SIZE BLOCK\_SIZE BLOCK\_SIZE B.width A.width

#### Tiled Matrix Multiply

```
global__ void mat_Mul(float* A, float* B, float* C, int ARows, int ACols, int BRows, int BCols, int CRows, int CCols) {
  float CValue = 0;
  int Row = blockldx.y*TILE_DIM + threadIdx.y;
  int Col = blockldx.x*TILE DIM + threadIdx.x;
  __shared__ float As[TILE_DIM][TILE_DIM];
  __shared__ float Bs[TILE_DIM][TILE_DIM];
  for (int k = 0; k < (TILE_DIM + ACols - 1)/TILE_DIM; k++) {
     if (k*TILE_DIM + threadIdx.x < ACols && Row < ARows) As[threadIdx.y][threadIdx.x] = A[Row*ACols + k*TILE_DIM +
threadIdx.x];
                                         As[threadIdx.y][threadIdx.x] = 0.0;
     else
     if (k*TILE_DIM + threadIdx.y < BRows && Col < BCols) Bs[threadIdx.y][threadIdx.x] = B[(k*TILE_DIM +
threadIdx.y)*BCols + Col];
                                         Bs[threadIdx.y][threadIdx.x] = 0.0;
     else
     __syncthreads();
     for (int n = 0; n < TILE DIM; ++n) CValue += As[threadIdx.y][n] * Bs[n][threadIdx.x];
     __syncthreads();
  if (Row < CRows && Col < CCols) C[((blockldx.y * blockDim.y + threadldx.y)*CCols)+(blockldx.x*blockDim.x)
+threadIdx.x]=CValue;
```

# Timing



http://www.umiacs.umd.edu/~ramani/cmsc828e\_gpusci/Lecture5.pdf

#### **BLAS** library

Basic Linear Algebra Subprograms

Quick Reference Guide: <a href="http://www.netlib.org/blas/blasqr.pdf">http://www.netlib.org/blas/blasqr.pdf</a>

http://www.netlib.org/blas/faq.html#8

# Matrix Multiply with CUBLAS

CUBLAS is the CUDA implementation of BLAS
(Basic Linear Algebra Subprograms)
<a href="http://docs.nvidia.com/cuda/cublas/">http://docs.nvidia.com/cuda/cublas/</a>

Consider scalars  $\alpha, \beta$ , vectors x, y, and matrices A, B, C.

Three levels of CUBLAS programs:

Level 1 BLAS 1: Scalar and vector based functions y → αx + y

Level 2 BLAS 2: Matrix-vector operations y → αAx + βy

Level 3 BLAS 3: Matrix-matrix operations C → αAB + βC

#### **BLAS**

Most routines have the following in their names:

- s single precision
- d double precision
- c complex single precision
- z complex double precision

#### Matrix types:

| general<br>symmetric | ge<br>sy |  |
|----------------------|----------|--|
| Hermitian            | he       |  |
| triangular           | tr       |  |

Example: dgemm is a double precision

general matrix multiply (Level 3)

# Legacy FORTRAN

NOTE:

BLAS originally written in FORTRAN. Arrays stored column major in FORTRAN, while they are row major in C++. Further, indexing starts at 1, rather than 0.

Need to have arrays in the correct format.

All arrays are flattened. From column major linear array to C++ 2D array:

encodes matrix B,

#define IDX2C(i,j,ld) (((j)\*(ld))+(i)) takes 2D array indices (i,j) into column major index where ld is leading dimension (# rows)

#### Check out IDX2C define for column major arrays

```
/* simple examination of IDX2C
#include <iostream>
#define IDX2C(i,j,ld) ((j)*(ld) + (i))
using namespace std;
my array is 1 2 3 4
            5 6 7 8
            9 10 11 12
*/
```

Note: Id = 3 leading dimension is the number of elements to count in linear array to get to the next column.

```
int main(){
 int a[12];
 cout<<endl <<"column major array\n";
 // define column major array using IDX2C
 // this is the way to use C++ for-loops to get column major linear array
 for (int i = 0; i < 3; i++) //rows
  \{for (int j = 0; j < 4; j++) //columns \}
    \{ a[IDX2C(i,i,3)] = i*4+i+1;
     cout<<"a["<<IDX2C(i,j,3)<<"]=" <<i*4+j+1<<",";
     cout<<"\n";
  cout<<"\n encoded linear array: "<<endl;
  for (int i=0; i<12; i++)
   cout<<a[i]<<", ";
   cout<<endl<<endl;
  //print out column major linear array in 2D form.
  for (int i = 0; i < 3; i++)
    \{for(int \ i = 0; \ i < 4; \ i++)\}
      \{cout << a[IDX2C(i,j,3)] << ", ";\}
      cout << endl;
```

#### Output: testIDX2C.cpp

```
my array is 1 2 3 4
             5 6 7 8
              9 10 11 12
column major array
a[0]=1,a[3]=2,a[6]=3,a[9]=4,
a[1]=5,a[4]=6,a[7]=7,a[10]=8,
a[2]=9,a[5]=10,a[8]=11,a[11]=12,
encoded linear array:
1, 5, 9, 2, 6, 10, 3, 7, 11, 4, 8, 12,
1, 2, 3, 4,
5, 6, 7, 8,
9, 10, 11, 12,
```

# Matrix Multiply with CUBLAS (kernel)

```
// Multiply the arrays A and B on GPU and save the result in C
// C(m,n) = A(m,k) * B(k,n)
void gpu_blas_mmul(const float *A, const float *B, float *C, const int m, const int k,
const int n) {
   int lda=m,ldb=k,ldc=m;
    const float alf = 1;
    const float bet = 0;
    const float *alpha = &alf;
    const float *beta = &bet;
    // Create a handle for CUBLAS
    cublasHandle_t handle;
    cublasCreate(&handle);
   // Do the actual multiplication
    cublasSgemm(handle, CUBLAS_OP_N, CUBLAS_OP_N, m, n, k, alpha, A, Ida,
B, Idb, beta, C, Idc);
   // Destroy the handle
    cublasDestroy(handle);
```

# From CUDA Samples

nvcc matrixMulCUBLAS.cpp -I/Developer/NVIDIA/CUDA-6.5/samples/common/inc - lcublas