



# 并行计算大作业

## ——并行优化矩阵乘法

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2019.12.27



# Outline

Outline

Introduction

Photo

Conclusion

References



# Backgrounds

- 11111111111111
- 22222222222222
- 33333333333333
- 44444444444444



# My Photo



图: hahahaha...



Emmm...

## Sequence Tagging Loss

$$\mathcal{L}_p = - \sum_{i=1}^S \sum_{j=1}^N p_{i,j} \log(\hat{p}_{i,j})$$

## Language Classifier Loss

$$\mathcal{L}_a = - \sum_{i=1}^S l_i \log(\hat{l}_i)$$

## Bidirectional Language Model Loss

$$\mathcal{L}_l = - \sum_{i=1}^S \sum_{j=1}^N \log(P(w_{j+1}|f_j)) + \log(P(w_{j-1}|b_j))$$



# References



Xuezhe Ma and Eduard Hovy. (2016).

## **End-to-end Sequence Labeling via Bi-directional LSTM-CNNs-CRF.**

In Proceedings of the 54th Annual Meeting of the Association for Computational Linguistics, pages 1064–1074, Berlin, Germany, August 7-12, 2016.



Marek Rei. (2017).

## **Semi-supervised Multitask Learning for Sequence Labeling.**

In Proceedings of the 55th Annual Meeting of the Association for Computational Linguistics, pages 2121–2130, Vancouver, Canada, July 30 - August 4, 2017.



# solving matrix mult on CUDA

- host vs device, distributed memory
- streamprocessors
- blocks, wraps, threads, shared memory

子任务：给定  $i, j$ , 计算  $C_{ij}$ , 也就是并行最外面的两个 for



# solving matrix mult on CUDA

- 轻量级线程，切换成本较低
- 基本可以平等划分任务
- 一个线程对应一个子任务（开  $n \times m$  个线程）







# shared memory

- 公式:  $C_{ij} = \sum_{k=0}^{bm} A_{ik} \cdot B_{kj}$
- 对于内层的两个循环  $j, k$ , 发现公用  $A_{ix}$
- 使用 share memory 缓存  $A$



# share memory

- block 内公用
- block\_size 必须是  $am$  的因子, 保证同块内使用相同的一行
- 别忘了 `__synchronize`



# Fast IO: buffered input & output

- `fread`, `fwrite` instead of `printf`, `scanf`
- no need to flush every time we read/write a number



# Occupancy: optimal size for the number of blocks and threads

- 以 wrap 为计算单元
- thread 中有 register
- block 中有 share memory
- 每个 multiprocessor 也有最大的 register 和 share memory 和 active threads
- 如果资源不够将无法派出所有的资源进行运算



# Occupancy: optimal size for the number of blocks and threads

- 我们需要让最大比例的显卡工作
- 引入 Occupancy 机制
- 通过 nvidia 官网上的 Occupancy calculator 进行运算
- `nvcc --ptxas-options=-v` to `nvcc` 查看内核函数使用寄存器个数
- 使用



# Bottleneck: IO

- GPU 计算矩阵瓶颈在于 IO
- 输出浮点数也会带来计算





# Bottleneck: IO

- 仿照 MPI 在 distributed memory 中使用异步 IO
- Host 一边传入 Device 的数据一边输出到文件
- Host 一边读入文件一遍 Copy 数据到 Device



# asynchronize IO: input

- `cudaStream_t`
- `cudaHostAlloc` instead of `malloc`
- `cudaMemcpyAsync` instead of `cudaMemcpy`







# code: buffered input

```
inline char next_char()
{
    static char buf[1000000], *p1 = buf,
               *p2 = buf;
    return p1 == p2 && (p2 = (p1 = buf)
        + fread(buf, 1, 1000000, fin), p1 == p2)
        ? EOF : *p1++;
}
```



# code: buffered output

```
inline void flush() {  
    fwrite(buffer, 1, s-buffer, stdout);  
    s = buffer;  
    fflush(stdout);  
}  
inline void print(const char ch) {  
    // putchar(ch); return;  
    if (s-buffer>OutputBufferSize-2) flush();  
    *s++ = ch;  
}
```



# code: share memory, block partition

```
__shared__ ld c_a[max_shared_size];  
int index = blockDim.x * blockIdx.x + threadIdx.x;  
if (index >= an * bm) return;  
int st = min(index, addi) * (workload+1) +  
max(0, index - addi) * workload, ed =  
st + workload + (index < addi ? 1 : 0);  
int shareda = min(am, max_shared_size);  
for (int p=st; p<ed; ++p) {  
    // ...  
}
```



# code: share memory, block partition

```
for (int p=st; p<ed; ++p) {  
    int i = p / bm, j = p % bm;  
    if (p % bm == 0) {  
        for (int j=0; j<shareda; ++j) {  
            c_a[j] = d_a[i * am + j];  
        }  
        __syncthreads();  
    }  
    ...  
}
```



# code: share memory, block partition

```
for (int p=st; p<ed; ++p) {  
    int i = p / bm, j = p % bm;  
    if (p % bm == 0) {  
        for (int j=0; j<shareda; ++j) {  
            c_a[j] = d_a[i * am + j];  
        }  
        __syncthreads();  
    }  
    ...  
}
```



# code: async IO, creating streams

```
int st = 0, ed = n * m;  
// printf("st=%d ed=%d, a=%p\n", st, ed, a);  
cudaStream_t stream[2];  
int mask = 0;  
cudaStreamCreate(&stream[0]);  
cudaStreamCreate(&stream[1]);  
int size;
```



# code: async IO, creating streams

```
cudaStream_t mainstream;
cudaStreamCreate(&mainstream);
// ...
copyMatrixAsync(h_a, d_a, an, am, mainstream);
// ... read h_b
copyMatrixAsync(h_b, d_b, bn, bm, mainstream);
// ...
matrixMult<<<grids, block_size, 0, mainstream>>>
(d_a, d_b, d_c, an, bm, am);
// ...
handleCudaError(cudaStreamSynchronize(mainstream));
// ...
outputMatrixAsync(h_c, d_c, n, m);
// pipeline: memcpy from device & output each rows
```



# code: async IO, pipelines

```
for (; st<ed; st+=size, mask^=1) {
    size = min(chunk_size, ed - st);
    handleCudaError(cudaMemcpyAsync(a + st,
        d_a + st, size * sizeof(ld),
        cudaMemcpyDeviceToHost, stream[mask]));
    // exit(0);
    if (st - chunk_size >= 0) {
        // printf("%d %d\n",st-chunk_size, st);
        handleCudaError(cudaStreamSynchronize(stream));
        outputinterval(a, st-chunk_size, st);
    }
}
```