### Randomization Algorithm to Compute Low-Rank Approximation

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# Outline

- Background
- Algorithm and Math Model
- Project Scheme

Done

To do

## Background

General SVD

$$A=U\sum V^t$$

matrices U =[u1u2...um] $\in$ Rm $\times$ m; V =[v1v2...vn] $\in$ Rn $\times$ n Σ=diag(σ1,...,σν),where Σ $\in$ Rm $\times$ n,v=min{m,n}and σ1  $\geq$ σ2  $\geq$ ... $\geq$ σν  $\geq$ 0.

Low-Rank SVD Approximation

$$A=U_k\sum_kV_k^t$$

LAPACK/MAGMA software framework

## Algorithm--Power iteration:

```
Matlab Code "svd_rand./" SVD approximation
function [u,s,v] = svd_rand(A, k, l, max_iters)
q = randn(n,k+l);
                                                        end
[q,r] = qr(q,0);
  for iter=1:(max_iters-1)
    p = A*q;
    q = A'*p;
    [q,r] = qr(q,0);
  end
```

```
p = A*q;
[p,b] = qr(p,0);
end
[x,s,y] = svd(b);
u_k = p*x(:,1:k);
s = s(1:k,1:k);
v_k = q*y(:,1:k);
```

## Algorithm and Math Model

Matrix	Size
Α	M-by-N
Q	N-by-(K+L)
P	M-by-(K+L)
В	(K+L)-by-(K+L)
X	(K+L)-by-(K+L)
ΥΤ	(K+L)-by-(K+L)
SI	(K+L)-by-1
S	K-by-1
$u_k$	M-by-K
$v_k$	N-by-K

Error= 
$$||A-U_KS_KV_K^T||_2$$
  
=  $(k+1)_{th}$  largest singular value of A

# Project Scheme

- 1. Implementing the randomized algorithm using LAPACK on CPU
- 2. Implementing the randomized algorithm using MAGMA on GPU
- 3. Implementing the out-of-memory randomized algorithm on GPU
  - -single queue
  - -UMA
  - -manual pipelining.
  - -Multiple GPUs using CUBLAS-XT
- 4. set up tester to compare performance
- 5.Application

#### Done

- 1. Implementing the randomized algorithm using LAPACK on CPU
- 2. Implementing the randomized algorithm using MAGMA on GPU
- 3. Implementing the out-of-memory randomized algorithm on GPU with single queue
- 4. set up tester to compare the performance

LAPACK & CPU&GPU&OUT-OF-MEMORY

## Out-of-Memory GPU Implementation

Device: Tesla K80, 823.5 MHz clock, 11439.9 MiB memory, capability 3.7

1 MiB =  $2^{20}$  bytes = 1024 kibibytes = 1048576bytes

11439.9Mib\*1048576=1.1996e+10 bytes

Sqrt(12e9/8)=3.8730e+04

# Out-of-Memory GPU Implementation

#### P=A\*Q

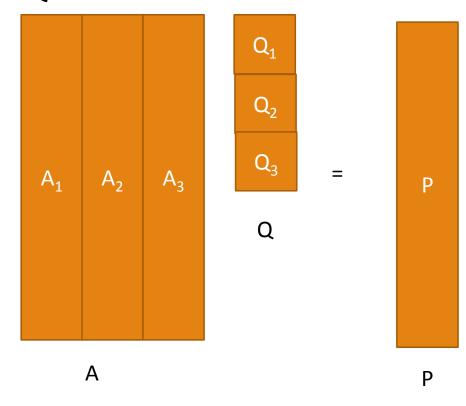
```
P=0;

For k=1,2,3.....

set (A_k to dA);

P=P+A_kQ_k;

end
```



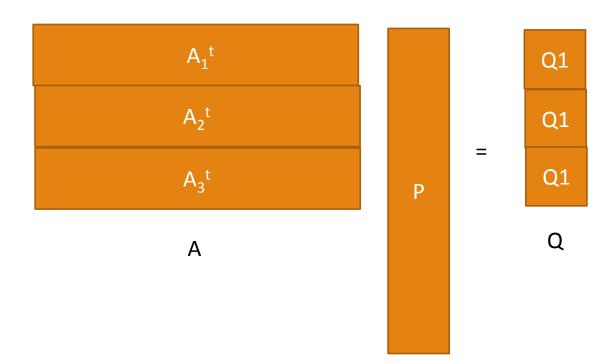
# Out-of-Memory GPU Implenmentation

#### $Q=A^{t*}P$

For 
$$k=1,2,3...$$
.

set  $(A_k \text{ to dA});$ 

$$Q_k=A_k^{t}P;$$
end



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#### Out-of-Memory GPU Implementation using single queue

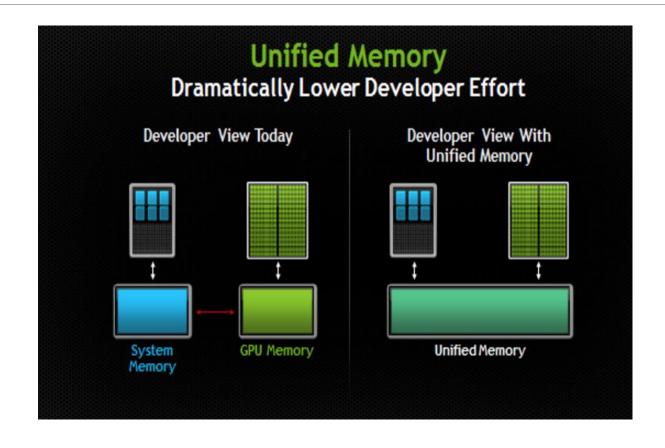


Time Line

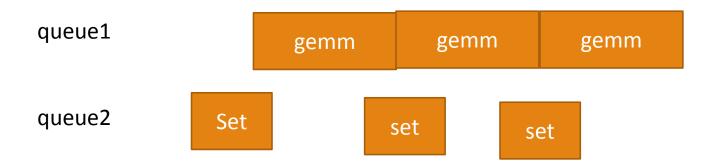
### To do

- Implementation on single GPU using UMA(Unified Memory Access)
- Implementation on single GPU using manually pipelining
- Implementation on multiple GPUs using CUBLAS-XT
- Application—image processing

## To do-Implementation on GPU using UMA



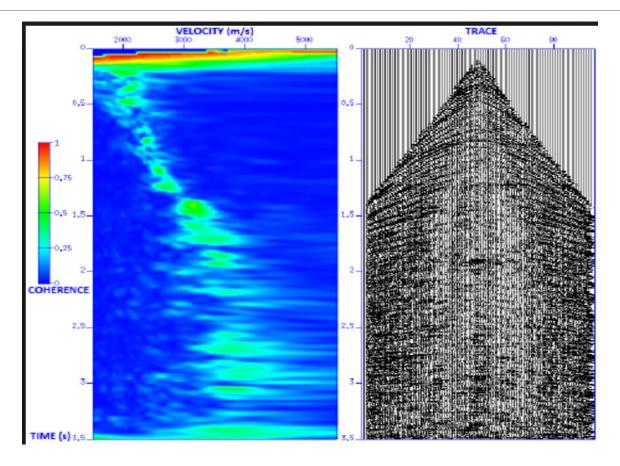
#### To do-Implementation on GPU using manually pipelining



Time Line

# To Do-Application

- Latent Semantic Indexing (LSI)
- Genetic clustering
- subspace tracking
- •image processing



### Reference

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