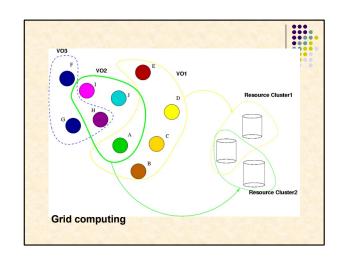
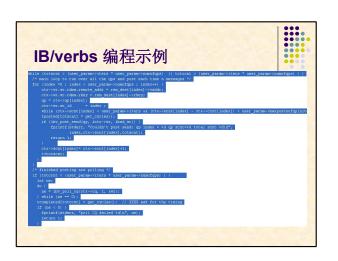
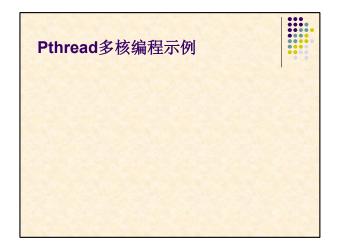


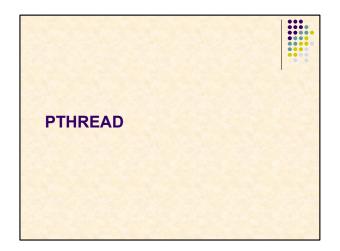
Ideas of parallel computation Multi-threading (e.g. Java) Concurrent processes (e.g. Unix) SIMD (e.g. Connection Machine-2) SPMD (e.g. MPI) MIMD (e.g. CORBA) Message-passing Memory-sharing Bulk-Synchronous Parallelism Work-flow task parallelism, data parallelism

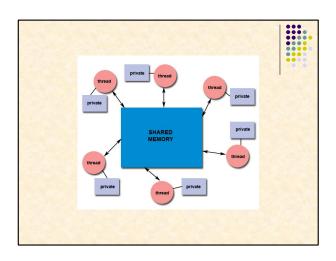


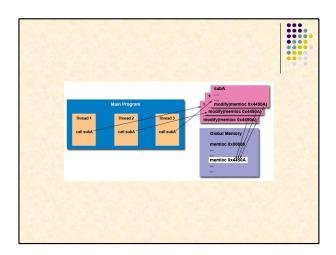


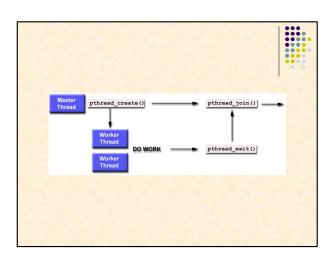






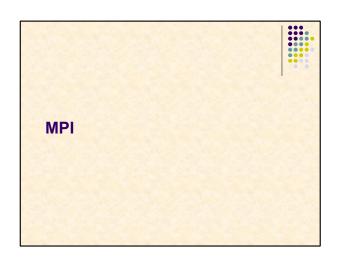






Thread 1	Thread 2	Balance	
Read balance: \$1000		\$1000	•
	Read balance: \$1000	\$1000	
	Deposit \$200	\$1000	
Deposit \$200		\$1000	
Update balance \$1000+\$200		\$1200	
	Update balance \$1000+\$200	\$1200	

```
#include <unistd.h> /* Symbolic Constants */
#include <sys/types.h> /* Primitive System Data Types */
#include <pthread.h> /* POSIX Threads */
#include <pthread.h> /* Semaphore */
void handler (void *ptr);
sem_t mutex;int counter; /* shared variable */
int main(){ int i[2];
pthread_t thread_a; pthread_t thread_b;
i[0] = 0; /* argument to threads */ i[1] = 1;
sem_init(&mutex, 0, 1); /* initialize mutex to 1 - semaphore */
pthread_create (&thread_a, NULL, (void *) &handler, (void *) &i[0]);
pthread_join(thread_b, NULL); pthread_join(thread_b, NULL);
sem_destroy(&mutex); /* destroy semaphore */
exit(0);} /* main() */
```



```
When the program is compiled and run with two processes, the output should be

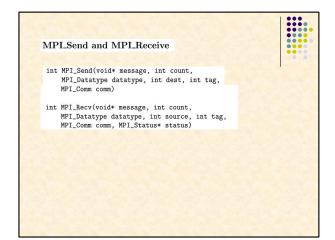
Greetings from process 1!

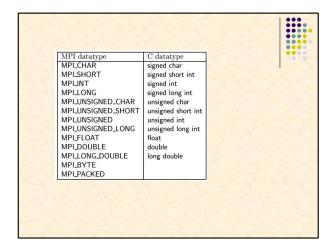
If it's run with four processes, the output should be

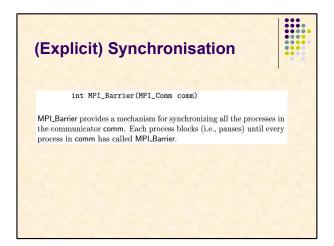
Greetings from process 1!

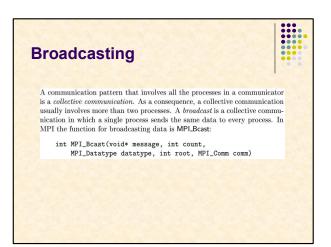
Greetings from process 2!

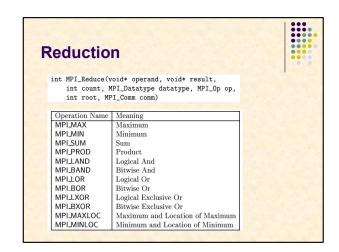
Greetings from process 3!
```

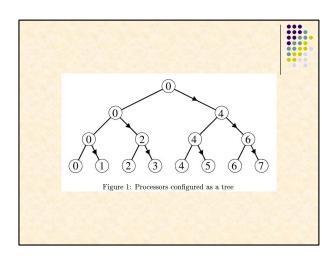




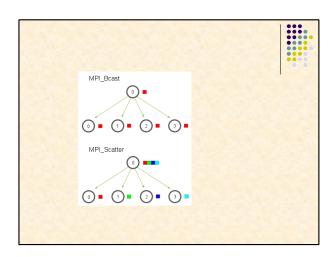


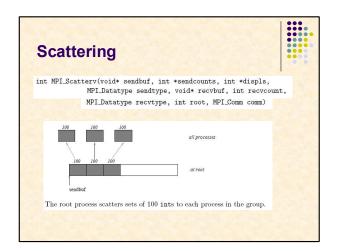


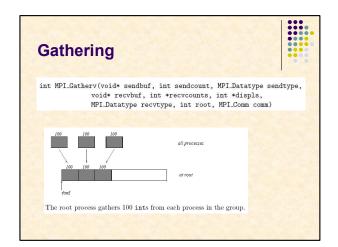


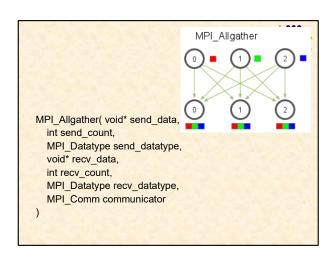


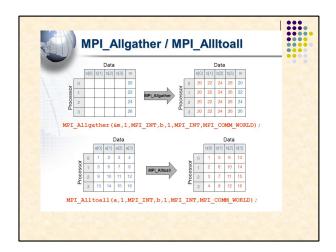
Advice to implementors. It is strongly recommended that MPLREDUCE be implemented so that the same result be obtained whenever the function is applied on the same arguments, appearing in the same order. Note that this may prevent optimizations that take advantage of the physical location of processors. (End of advice to implementors.)

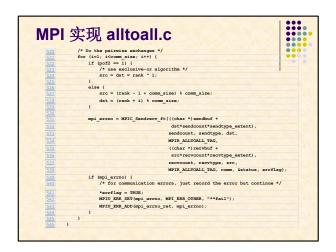












```
Communication Groups

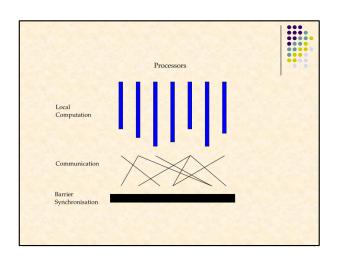
MPI_Group MPI_GROUP_WORLD;
MPI_Group first_row_group;
MPI_Comm first_row_comm;
int row_size;
int* process_ranks;

/* Make a list of the processes in the new
* communicator */
process_ranks = (int*) malloc(q*sizeof(int));
for (proc = 0; proc < q; proc++)
    process_ranks[proc] = proc;

/* Get the group underlying MPI_COMM_WORLD */
MPI_Comm_group(MPI_COMM_WORLD, &MPI_GROUP_WORLD);

/* Create the new group */
MPI_Group_incl(MPI_GROUP_WORLD, q, process_ranks, &first_row_group);

MPI_Comm_create(MPI_COMM_WORLD, first_row_group, &first_row_comm);
```

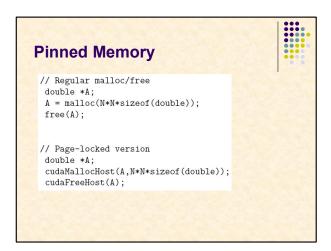


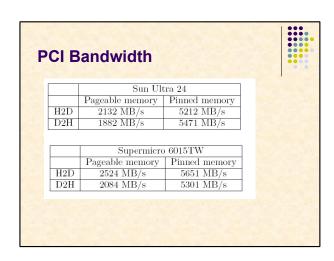


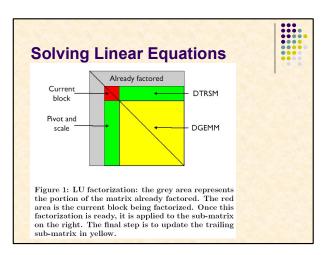
LINPACK (wiki)

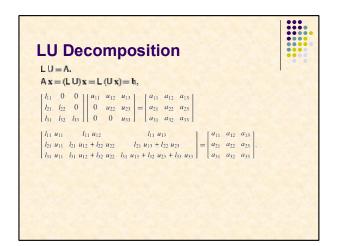
- The LINPACK Benchmarks are a measure of a system's floating point computing power. Introduced by Jack Dongarra, they measure how fast a computer solves a dense N by N system of linear equations Ax = b, which is a common task in engineering. The solution is obtained by Gaussian elimination with partial pivoting, with 2/3·N³ + 2·N² floating point operations. The result is reported in millions of floating point operations per second.
- Massimiliano Fatica, Accelerating linpack with CUDA on heterogenous clusters, GPGPU'09.
 Slides extracted from the above paper

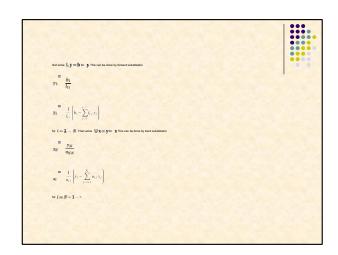
SUN Ultra 24 workstation with an Intel Core2 Extreme Q8550 (3.0GHz) CPU and 8GB of memory plus a Tesla C1060 card. Cluster with 8 nodes, each node connected to half of a Tesla S1070 system, containing 4 GPUs, so that each node is connected to 2 GPUs. Each node has 2 Intel Xeon E5462 (2.8GHz with 1600Mhz FSB) and 16GB of memory. The nodes are connected with SDR (Single Data Rate) Infiniband.

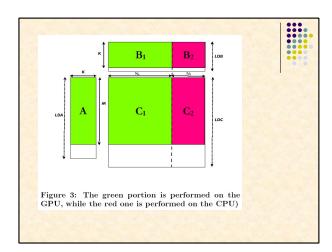


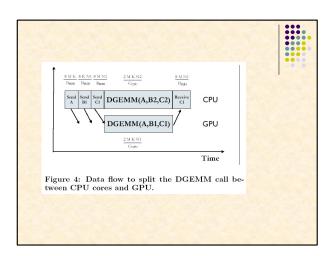












 B_{H2D} : Bandwidth from host to device expressed in GB/s G_{GPU} : Sustained performance of DGEMM on the GPU expressed in GFlops G_{CPU} : Sustained performance of DGEMM on the CPU expressed in GFlops B_{D2H} : Bandwidth from device to host expressed in GB/sA DGEMM call on the host CPU performs 2KMN operations, so if the CPU cores can perform this operation at G_{CPU} the total time is: $T_{CPU}(M,K,N) = 2\frac{MKN}{G_{CPU}}$ The total time to offload a DGEMM call to the GPU has an I/O component that accounts for both the data transfer from the CPU memory space to the GPU memory space and vice versa plus a computational part once the data is on the GPU. We can express this time as: $T_{GPU}(M,K,N) = 8\frac{(MK + KN + MN)}{B_{H2D}} + 2\frac{MKN}{G_{GPU}} + 8\frac{(MN)}{B_{D2H}}$ the factor 8 is the size of a double in bytes.
The optimal split will be $T_{CPU}(M,K,N2) = T_{GPU}(M,K,N1) \quad \text{with} \quad N = N1 + N2$

For an initial approximation of the optimal split fraction $\eta=N1/N$, we can omit the transfer time $(O(N^2))$ compared to the computation $(O(N^3))$. From a simple manipulation, the optimal split is $\eta=\frac{G_{GPU}}{G_{GPU}+G_{CPU}}$ On the cluster, where the quad core Xeon has a DGEMM performance of 40 GFlops and the GPU a DGEMM performance of 82 GFlops, this formula predicts $\eta=0.67$, very close to the optimal value of 0.68 found by experiments.

```
// Copy A from CPU memory to GPU memory devA
status = cublasSetMatrix (n, k , sizeof(A[0]), A, lda, devA, m_gpu);

// Copy Bi from CPU memory to GPU memory devB
status = cublasSetMatrix (a, n_gpu, sizeof(B[0]), B, ldb, devB, k_gpu);

// Copy Cl from CPU memory to GPU memory devC
status = cublasSetMatrix (a, n_gpu, sizeof(C[0]), C, ldc, devC, m_gpu);

// Perform DGEMM(devA,devB,devC) on GPU
// Control immediately return to CPU
cublasSemm('n', 'n', m, n_gpu, k, alpha, devA, m,devB, k, beta, devC, m);

// Perform DGEMM(A,B2,C2) on CPU
dgemm_cpu('n', 'n', m,n_cpu,k, alpha, A, lda,B+ldb*n_gpu, ldb, beta,C*ldc*n_gpu, ldc);

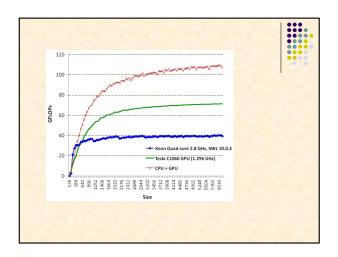
// Copy devC from GPU memory to CPU memory C1
status = cublasGetMatrix (m, n, sizeof(C[0]), devC, m, C, *ldc);
```

It turns out that on Intel systems using Front Side Bus (FSB), it is better not to overlap the transfer to the GPU with computations on the CPU (the memory system cannot supply data to both the PCIe and the CPU at maximum speed).

The DGEMM function call in CUBLAS maps to several different kernels depending on the size of the matrices. The best performance is achieved when M is multiple of 64 and K and N multiple of 16. Performance numbers for different choices of M, K and N are shown in table 5. When all the above conditions are satisfied, the GPU can achieve 82.4 Gflops, 95% of the peak double precision performance.

M	K	N	M%64	K%16	N%16	Gflops
448	400	12320	Y	Y	Y	82.4
12320	400	1600	N	Y	Y	72.2
12320	300	448	N	N	Y	55.9
12320	300	300	N	N	N	55.9

Table 3: DGEMM performance on the Tesla S1070 GPU (1.44 GHz) with data resident in GPU memory



The peak performance of the CPU is 48 GFlops, a CPU-only Linpack sustains performance in the 70%-80% range for very large problem sizes. The CUDA accelerated version with pinned memory (limited today to 4GB per MPI process) delivers 83.2 Gflops, 66% of the combined 125 GFlops system peak performance (the Tesla C1060 has a peak performance of 77.7 Gflops). The accelerated solution also delivers good performance for small problem sizes and is not very sensitive to the size of NB.

	T/V	N	NB	P	Q	Time(s)	Gflops
ı	WR00L2L2	23040	960	1	1	97.91	83.28
İ	WR00L2L2	7432	960	1	1	5.47	50.01
ı	WR00L2L2	7432	1152	1	1	5.47	53.56

Table 4: Linpack performance on a Sun Ultra 24 workstation with 1 Intel(R) Core2 Extreme Q6850 (3.0Ghz) and a Tesla C1060 using HPL with pinned memory for different N and NBs.

Table 5 shows the same results using pageable memory. The use of pinned vs. regular pageable memory increases the performance from 70 to 84 Gflops.

T/V	N	NB	P	Q	Time(s)	Gflops
WR00L2L2	23040	960	1	1	117.06	69.66
WR00L2L2	23040	1152	1	1	117.06	70.64
WR00L2L2	7432	960	1	1	6.09	44.94
WR00L2L2	7432	1152	1	1	5.97	45.86

Table 5: Linpack performance on a Sun Ultra 24 workstation with 1 Intel(R) Core2 Extreme Q6850 (3.0Ghz) and a Tesla C1060 using HPL with pagable memory

On the workstation, the biggest problem that can be solved with the available memory is N=32320 and the Linpack score is now 90 Gflops, 72% of peak performance.

Sun Ultra 24							
T/V	N	NB	P	Q	Time(s)	Gflops	
WR00R2L2	32320	1152	1	1	250.01	90	
Cluster							
T/V	N	NB	P	Q	Time(s)	Gflops	
WR11R2L2	118114	960	4	4	874	1258	

Table 8: Linpack performance using HPL with pinned memory (pre-release CUDA version)