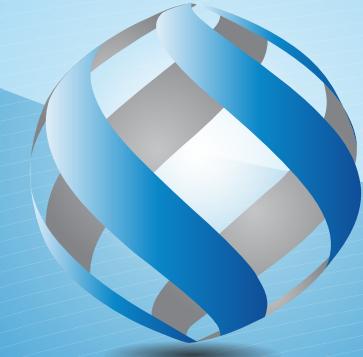
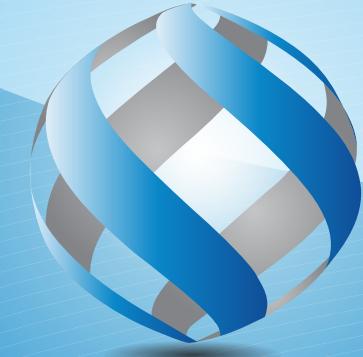


CUDA-enabled Math Libraries Overview



NVIDIA Libraries

- <http://developer.nvidia.com/cuda-tools-ecosystem>
- CUDA Toolkit
 - CUBLAS – dense Linear Algebra
 - CUSPARSE – sparse Linear Algebra
 - CUFFT – Fast Fourier Transforms
 - CURAND – random numbers generators
- Also:
 - Thrust – C++ template library
 - NPP – image & signal processing



3rd party Libraries

- MAGMA – dense Linear Algebra, an open-sourced equivalent to CUBLAS with extensions
- CUSP – sparse Linear Algebra and iterative solvers
- OpenCurrent – PDE solvers for regular grid



CUBLAS

- Implements BLAS program interface
- Multidimensional arrays allocated in column-major order (fortran)

Level	Computational complexity	Functions examples
1 (vector operations)	$O(N)$	$\text{AXPY: } y := \alpha x + y$ $\text{DOT: } \text{dot} := (x,y)$
2 (matrix-vector)	$O(N^2)$	GEMV – multiplate a general matrix-vector
3 (matrix-matrix)	$O(N^3)$	GEMM – multiplate two general matrices

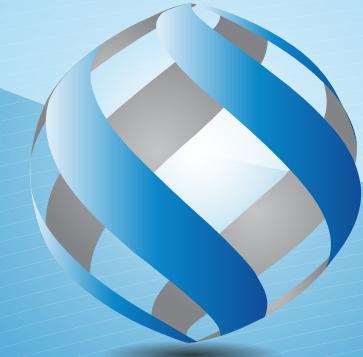
- Documentation is shipped together with CUDA toolkit (CUBLAS_Library.pdf)



CUBLAS

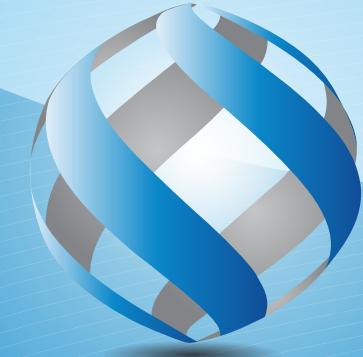
- ➊ Naming rule: cublas<T><func>
 - <T> - data type {F, D, C, Z} corresponds to real, single and double precision; complex, single and double precision
 - <func> - 3-4 letter of classic name of BLASfunction
 - Example: cublasDgemm

- ➋ In API 'v2' (available since CUDA 4.0) handles are used to make library thread-safe (when using several cuda Streams or multiple GPU)



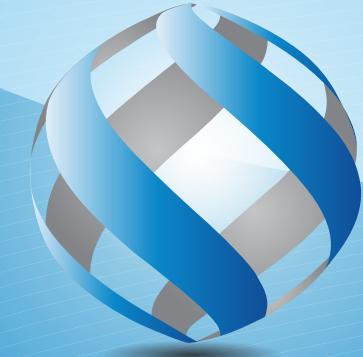
CUBLAS - pipeline

- ➊ Initialize CUBLAS handle using *cublasCreate* function
- ➋ Allocate required memory on GPU and load the input data
- ➌ Call the required CUBLAS function sequence
- ➍ Transfer the results to host memory
- ➎ Free CUBLAS handle using *cublasDestroy* function



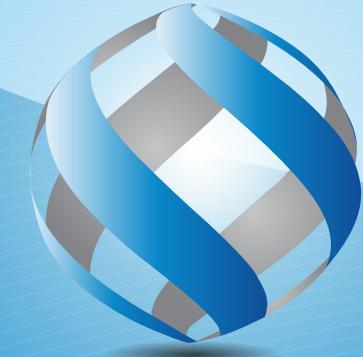
CUSPARSE

- **Purpose:** sparse matrix & vectors operations
- Sparse = significant amount of zero elements →
dense representation is ineffective, indexed
formats are used instead:
 - Sparse vector: 2 arrays – indexes and values
 - Sparse matrix: Coordinate Format (COO), Compressed
Sparse Row (Column) Format (CSR, CSC)



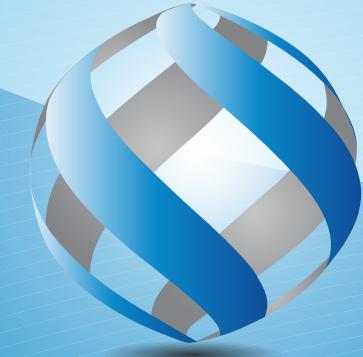
CUSPARSE – features

- ➊ 4 groups of functions: `cusparse<T><func>`
 - On sparse vectors and dense vectors
 - On sparse matrices and dense vectors
 - On sparse vectors and multiple dense vectors
 - Matrix Format conversions
- ➋ Supports real and complex data of single and double precision



CUSPARSE – pipeline

- 1) Allocate memory and initialize data on GPU
- 2) Initialize the CUSPARSE library
- 3) Perform required CUSPARSE operation
- 4) Release memory and operation handle



CURAND

- ➊ Library of pseudo- and quasi- random values generators
- ➋ Based on XORWOW and Sobol algorithms
- ➌ Available distributions: uniform, normal, lognormal
- ➍ Has two interfaces:
 - For host (generation on CPU or on GPU)
 - For device (generation on GPU)



CURAND – pipeline (host)

- ➊ Create generator using *curandCreateGenerator()*
- ➋ Set properties of generator (for example, initial state:
curandSetPseudoRandomGeneratorSeed())
- ➌ Allocate the GPU memory using *cudaMalloc()*
- ➍ Run the generator, using the appropriate distribution
- ➎ Use the results of generator
- ➏ If necessary, generate additional data
- ➐ Destroy generator *curandDestroyGenerator()*

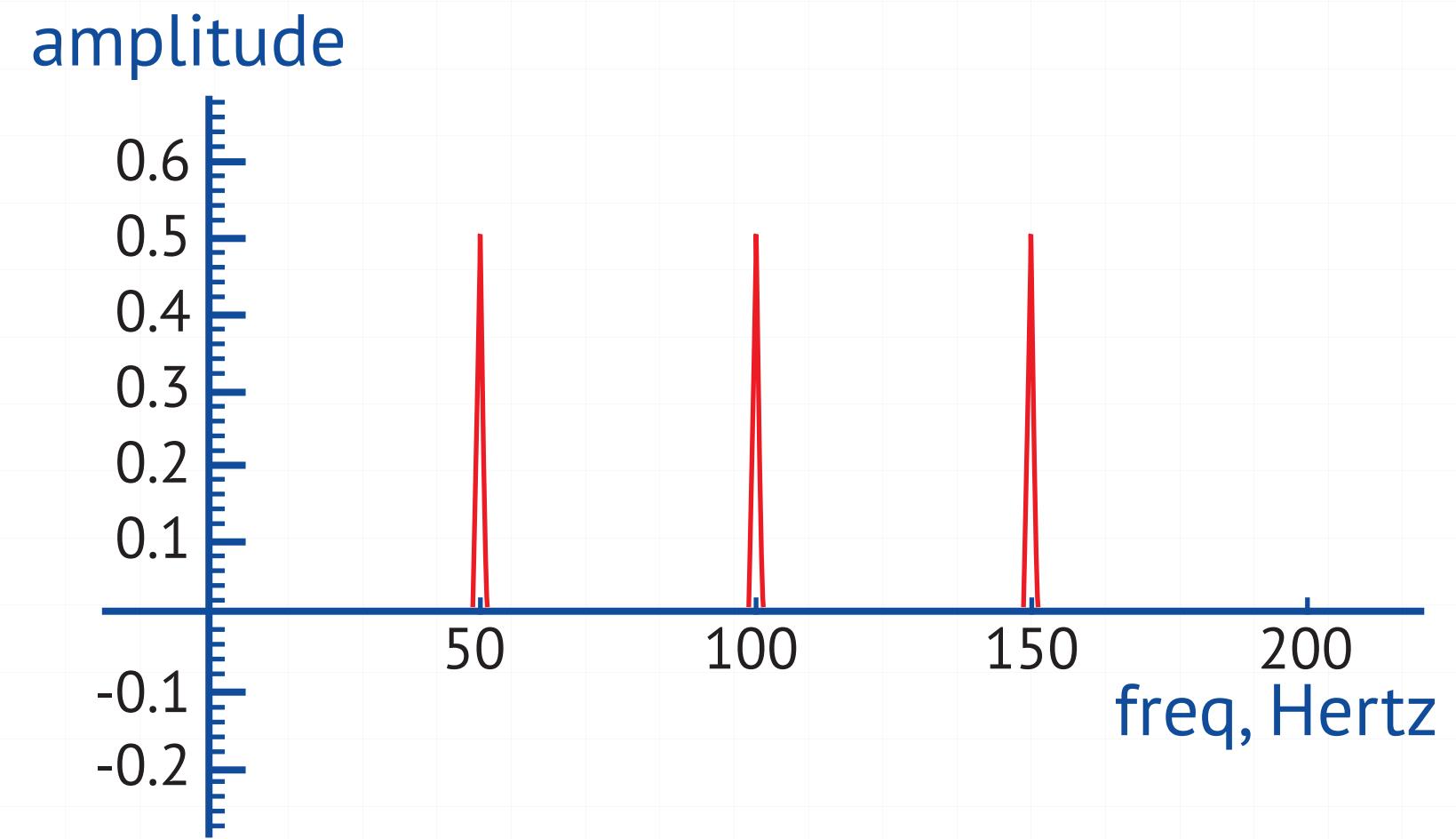
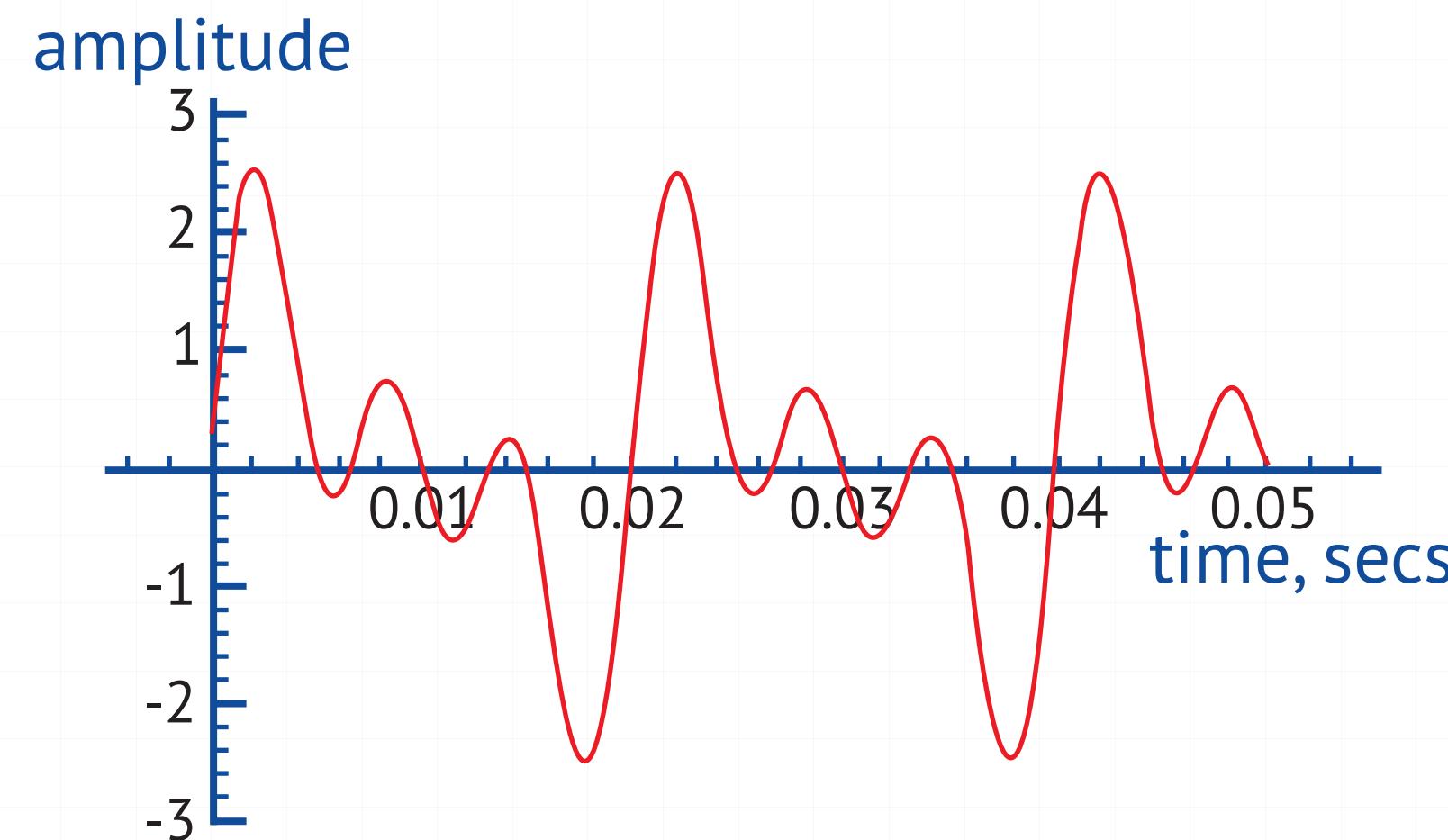


CURAND – pipeline (device)

- ➊ Allocate the array of states (*curandState*) of generators in each thread
- ➋ Create and run the kernel, initializing the *curandState* for each thread
- ➌ Create and run the kernel, using random values, generated by *curand(curandState)*
- ➍ Free memory

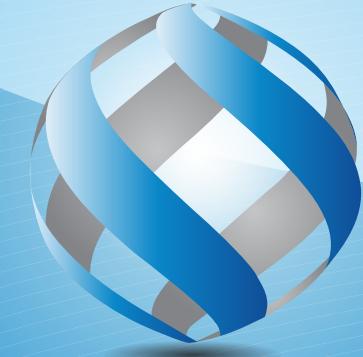


CUFFT – the Fast Fourier Transforms



$$F(x) = \sum_{n=0}^{N-1} f(n) e^{-j2\pi \left(x \frac{n}{N} \right)}$$

$$f(n) = \frac{1}{N} \sum_{x=0}^{N-1} F(x) e^{j2\pi \left(x \frac{n}{N} \right)}$$



CUFFT – the Fast Fourier Transforms

- ➊ The interface is similar to FFTW, supports FFTW compatibility mode
- ➋ Implements Cooley–Tukey and Bluestein's FFT algorithms
- ➌ 1-d, 2-d and 3d real and complex transforms of single or double precision
- ➍ Supports asynchronous transforms with CUDA streams
- ➎ Unnormalized output: $\text{IFFT}(\text{FFT}(A)) = \text{len}(A)^* A$



CUFFT – the pipeline

- 1) Allocate memory on GPU
- 2) Create FFT transform plan with the required size and type
- 3) Perform FFT transforms on data in GPU memory
- 4) Delete transform plan and release the GPU memory



CUFFT – sample



Consider the Poisson equation:

$$\begin{cases} \Delta u(p) = \rho(p), p \in \Omega = (x, y), 0 \leq x, y \leq 1, \\ \rho(x, y) = \frac{s(x, y) - 2\sigma^2}{\sigma^4} \exp^{-\frac{s(x,y)}{2\sigma^2}}. \end{cases}$$



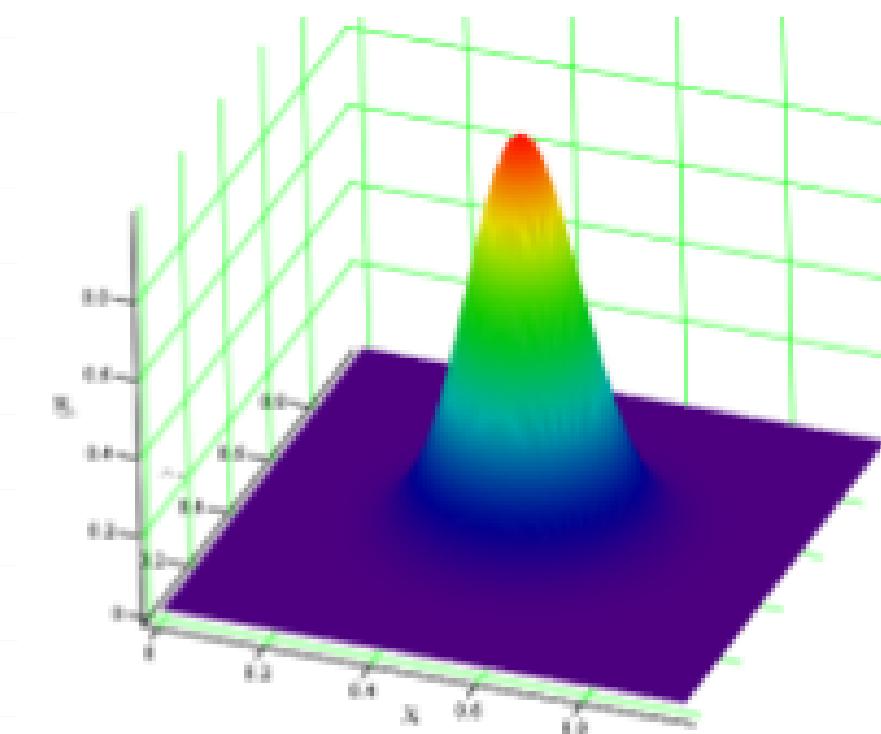
The known exact solution is: $u_0(x, y) = \exp^{\frac{s(x,y)}{2\sigma^2}}$



With some assumptions the approximate numerical solution can be found:

$$\bar{u}(n, m) = \bar{\rho}(n, m) h^2 (W^{-n} + W^n + W^{-m} + W^m - 4)^{-1}, \quad \bar{\rho} = \frac{1}{N^2} \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} \rho(x_k, y_j) W^{nk+mj}$$

$$u(x_k, y_j) = \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} \bar{u}(n, m) W^{nk+mj}.$$



$$W = \exp^{i \frac{2\pi}{N}}$$