

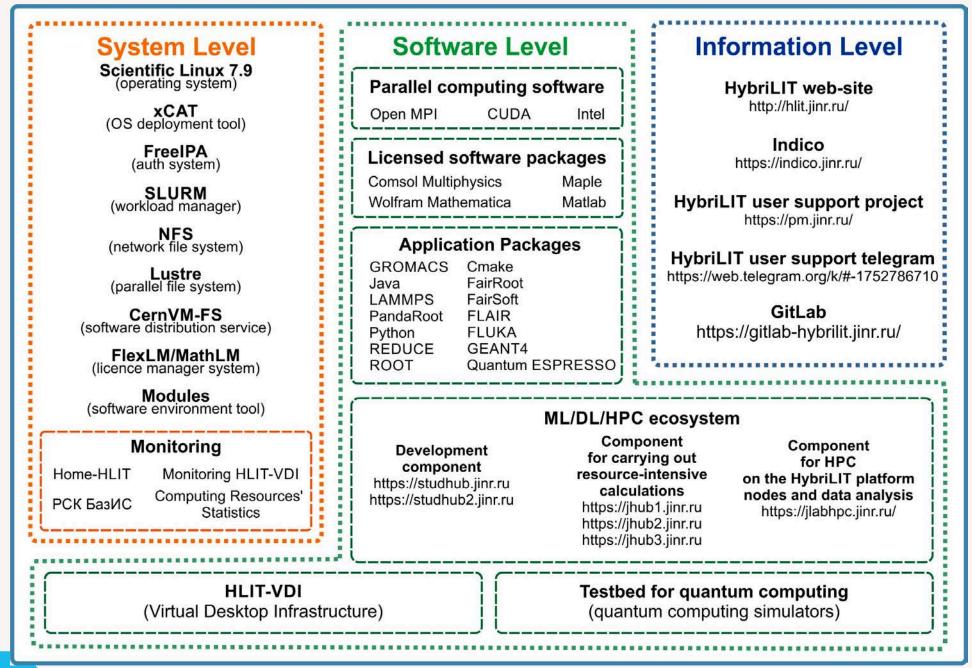
# Capabilities of the heterogeneous platform HybriLIT for quantum computing

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### Heterogeneous platform HybriLIT



### **Development of the heterogeneous platform HybriLIT**

Cluster HybriLIT **2014**: Peak performance: **50 TFLOPS** double precision **140 TFLOPS** single precision #18 in Top50 Supercomputer "Govorun" First stage 2018 Peak performance: 500 TFLOPS double precision 1 PFLOPS single precision #10 in Top50 Supercomputer "Gorvorun" Second stage 2019 Peak performance: 860 TFLOPS double precision 1.7 PFLOPF single precision 288 TB UDSS with I/O >300 Gb/s

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### Supercomputer "Govorun"

### **CPU component**

- 21x Servers with Intel Xeon Phi Intel Xeon Phi 7290 (72 cores @1.50 GHz), 96 GB RAM
- 76x Servers with Intel Xeon Scalable Gen2 (RSC Tornado TDN511) 2x Intel Xeon Platinum 8268 (24 Cores @2.90 GHz), 192 GB RAM
- 32x Servers with Intel Xeon Scalable Gen2 (RSC Tornado TDN511S) 2x Intel Xeon Platinum 8368Q (38 Cores @2.60 GHz), 2 TB RAM

### **GPU component**

- 5x Servers NVIDIA V100

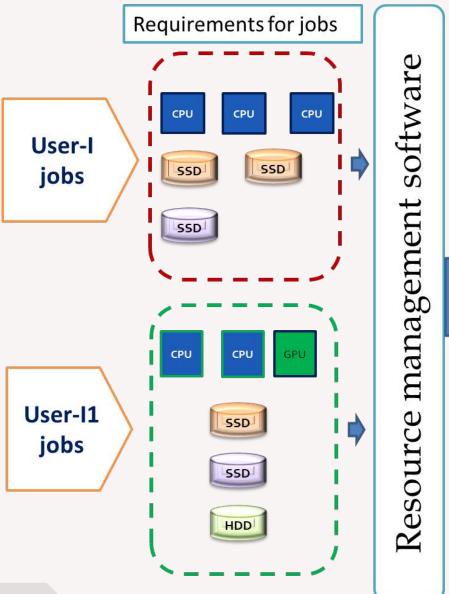
2x Intel Xeon E5-2698 v4 (20 cores @2.20 GHz), 8x NVIDIA V100 16 GB, 512 GB RAM

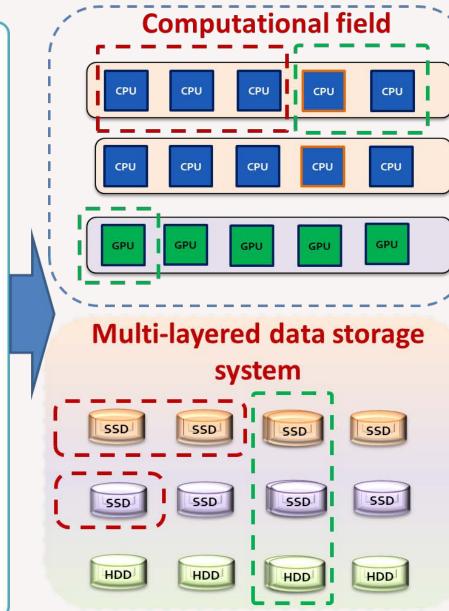
- 5x Servers with NVIDIA A100 2x AMD EPYC 7763 (64 Cores @2.45 GHz), 8x NVIDIA A100 80 GB, 2 TB RAM

Data storage system: 8.6 PB

Total peak performance 1.7 PFLOPS double precision 3.4 PFLOPS single precision

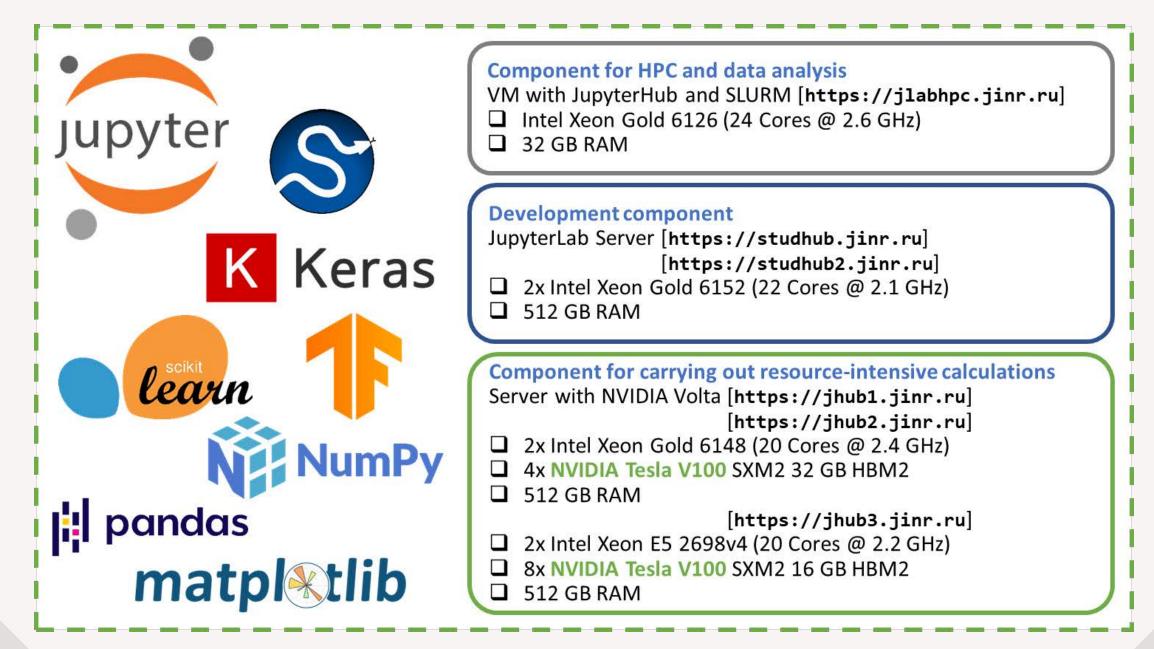
## **Orchestration and hyperconvergence**





The SC "Govorun" has unique properties for the flexibility of customizing the user's job. For his job the user can allocate the required number and type of computing nodes and the required volume and type of data storage systems. This property enables the effective solution of different tasks, which makes the SC "Govorun" a unique tool for research underway at JINR.

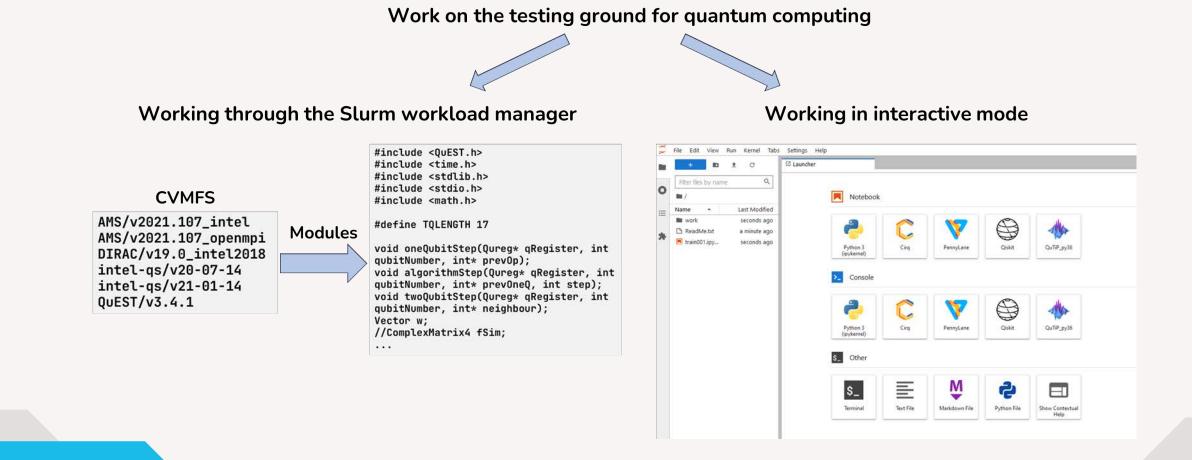
### Ecosystem for tasks of machine learning, deep learning and data analysis



# **Quantum simulators**

While quantum computers are not available for widespread use, various simulators of quantum computing on classical computers are being developed.

These are libraries on various programming languages or frameworks that allow to create, transform, optimize and effectively simulate quantum circuits. So, they allow user to completely control the behavior of a quantum system.



# Testbed for quantum computing. Working through the workload manager.

### The main advantages:

- the ability to perform multi-node computations using MPI technology;
- the use of resources of the entire heterogeneous platform.

### The order of the user's work:

- connection to the HybriLIT heterogeneous platform;
- setting the necessary environment variables for each individual quantum simulator in a work session using the Environment Modules. All available simulators are installed in the CVMFS network file system;
- preparation of a quantum algorithm;
- writing a script-file for run the task: the necessary computing resources (CPU, GPU, RAM), computation time;
- running a task through the SLURM;
- after the program is completed, the user can view the results of the computations and data about the process of the program in output files in his working directory.

# Test task. QuEST simulator.

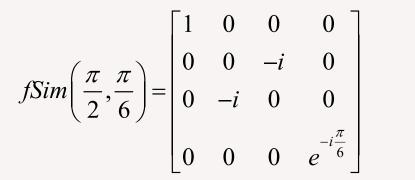
https://quest.qtechtheory.org

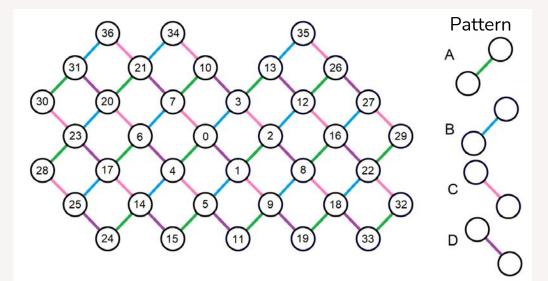
A task using the QuEST simulator together with Ekaterina Kotkova was prepared.

# Quantum algorithm for creating a randomized quantum scheme

- 1. **m** steps of the algorithm consisting of two parts:
  - a. Applying one-qubit gates to all qubits, which are randomly selected from the set  $\{\sqrt{X}, \sqrt{Y}, \sqrt{W}\}$ , where  $W = (X + Y)/\sqrt{2}$ .
  - b. Depending on the step number, applying two-qubit gates according to the pattern **ABCDCDAB**: at the first step, gates are applied between qubits with numbers corresponding to the pattern **A**, on the second, pattern **B**, etc.
- 2. Repeat of step 1.
- 3. Measurement of all qubits.

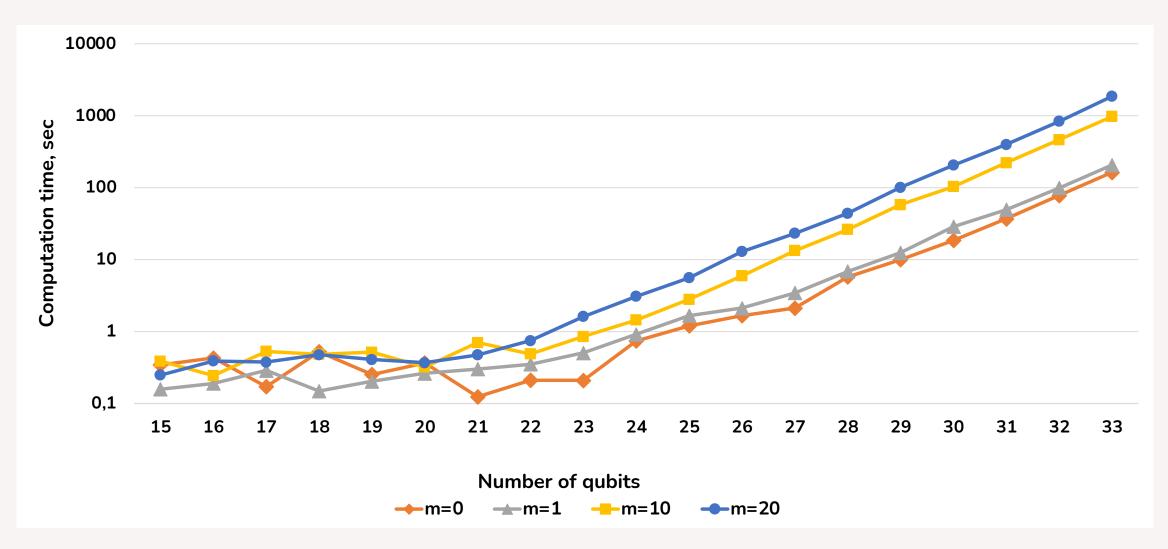
Two-qubit gate:





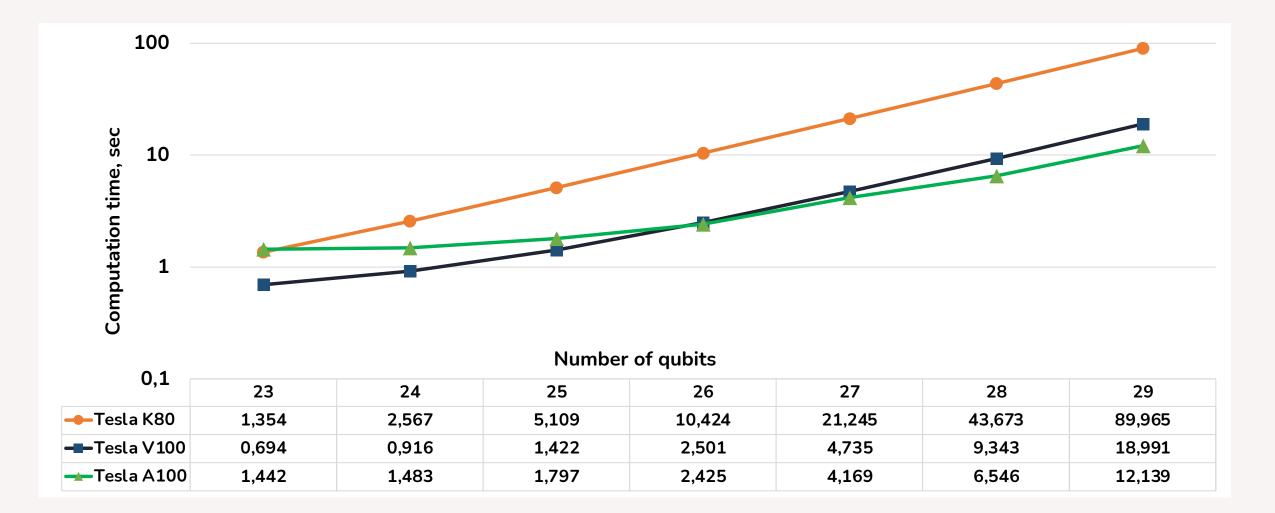
## QuEST simulator. Computations on CPU. Execution time.

Server specification: 2x Intel Xeon Platinum 8368Q (38 cores @ 2.60 GHz), 2 TB RAM



Computation time of the task depending on the number of qubits for different numbers of threads *m* 

### QuEST simulator. Computations on GPU. Execution time.



Computation time of the task depending on the number of qubits on different GPUs

# QuEST simulator. Computations on GPU. Memory usage.

GPU: NVIDIA Tesla V100

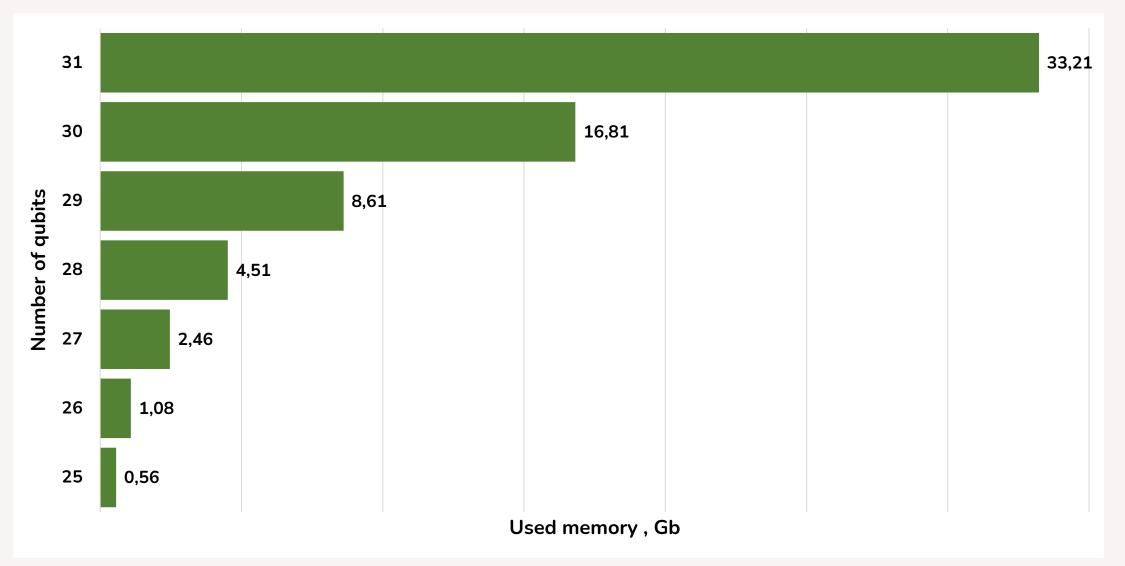


Diagram of dependency used GPU memory on the number of qubits

# Testbed for quantum computing. Working in interactive mode.

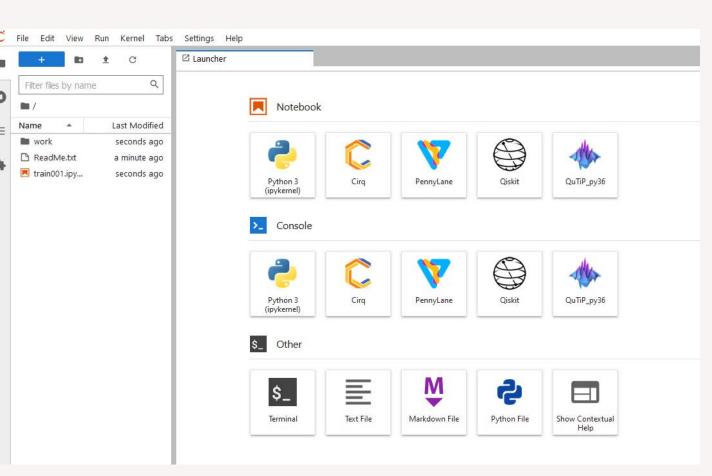
### The main advantages:

- the ability to visually develop algorithms, visualize quantum circuits;

- available Python language materials can significantly speed up research.

### Technical implementation.

To work with quantum simulators, a separate server has been allocated, on which the Anaconda package is locally installed. Quantum simulators are installed in virtual environments. Due to this, it is possible to avoid conflicts between versions of libraries that are installed with simulators. Virtual environments are output to the JupyterLab interface by creating a computing core in an interactive ipython shell, which is installed in each environment separately.



#### Servers specification:

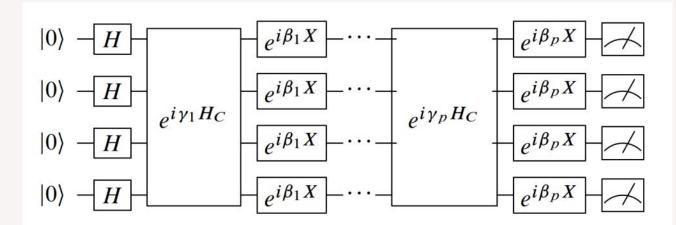
- 2x Intel Xeon E5-2698 (20 cores @ 2.2 GHz), 512 GB RAM, 8x NVIDIA Tesla V100 16 GB - 2x AMD EPYC 7763 (64 cores @ 2.45 GHz), 2 TB RAM, 8x NVIDIA Tesla A100 80 GB

# Task 2. Searching for the state with the lowest energy in the Ising model with a longitudinal magnetic field using the quantum approximation optimization algorithm (QAOA).

In QAOA, the function of a quantum computer is to construct a variational ansatz  $|\Psi(\beta, \gamma)\rangle$  of a wave function with parameters  $\beta = (\beta_1, ..., \beta_n)$  and  $\gamma = (\gamma_1, ..., \gamma_n)$  and measure the quantum energy averages  $\mathcal{E}(\beta, \gamma)$  as the average for the Hamiltonian  $\mathcal{H}$ :

 $\mathcal{E}(\beta, \gamma) = \langle \psi(\beta, \gamma) | \mathcal{H} | \psi(\beta, \gamma) \rangle.$ On a classic computer, the process of optimizing parameters takes place to reash a minimum value of the average  $\mathcal{E}(\beta, \gamma)$ .

The solution to the problem is to find a pair of parameters  $\beta$ ,  $\gamma$  at which the energy value  $\mathcal{E}(\beta, \gamma)$  will be minimal.



A quantum circuit to the variation ansatz of QAOA

$$\langle \psi(\gamma,\beta)\rangle = \underbrace{U(\beta_p,B)U(\gamma_p,\mathcal{H})}_{p} \cdots \underbrace{U(\beta_1,B)U(\gamma_1,\mathcal{H})}_{1} H^{\otimes n}|0\rangle^{\otimes n}$$

The problem statement is presented in detail in the work Yu. Palii, A. Bogolubskaya, D. Yanovich. Quantum approximation optimization algorithm for the Ising model in an external magnetic field https://indico.jinr.ru/event/3505/contributions/21552

## Software implementation. The Cirq library.

https://quantumai.google/cirq

- The software implementation of the algorithm is performed using the Cirq library.
- The qsim optimized simulator is integrated into the Cirq library. It is written in C++ and uses SIMD instructions for vectorization, OpenMP for CPU computations and CUDA for GPU computations.

The use of various parallelization technologies is set by the command **qsimcirq.QSimOptions()** 

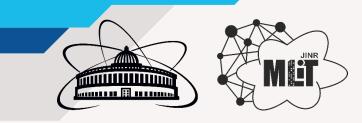
```
which sets the following basic parameters:
cpu_threads: int = XXX – number of OpenMP threads,
use_gpu: bool = False/True – use CUDA or NVIDIA cuStateVec,
gpu_mode: int = 0/1 – use CPU or GPU in the simulation process.
```

cuStateVec is a library from the NVIDIA cuQuantum SDK for modeling state vectors on the GPU.

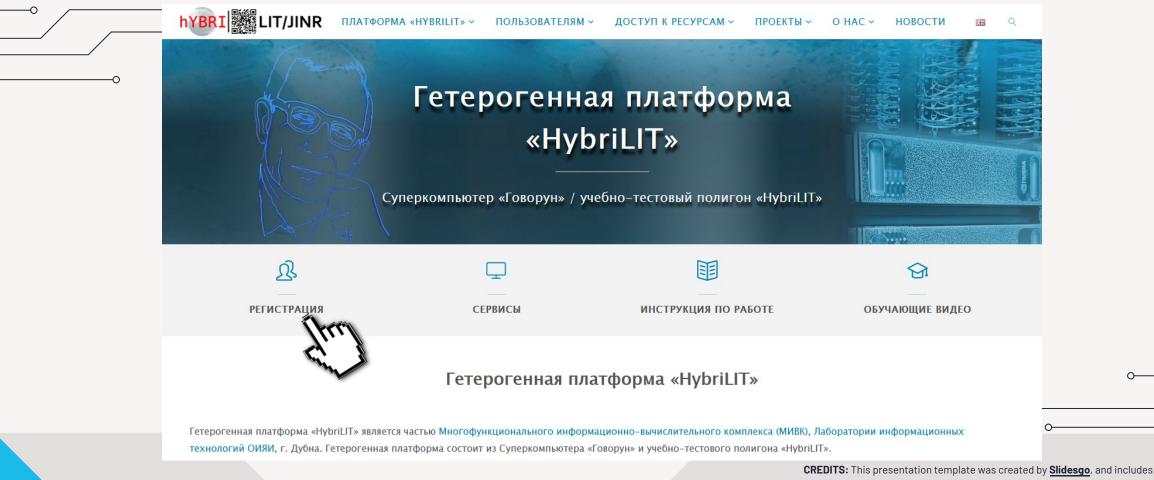
Ising Model 3x3x3 lattice	AMD EPYC 7763,	Intel Xeon Platinum	NVIDIA A100,
27 qubits	128 threads	8368Q, 128 threads	cuStateVec
Computation time	3 h 20 min	3 h 10 min	14 min 35 sec

# Conclusion

- A quantum computing polygon has been organized on the resources of the heterogeneous platform HybriLIT. It is provides the opportunity to work with quantum simulators in two access modes: through the workload manager and in interactive mode.
- The heterogeneous structure of the platform allows to quickly change the characteristics of the testbed, adjusting it to the user tasks requirements, adding servers with the necessary computing components, both CPU and GPU.
- The ability to simulate systems with different numbers of qubits depends on the architecture of the simulator. On the considered tasks, 33 qubits were simulated on the QuEST simulator, and 27 qubits on the Cirq simulator.
- Computations for Ising Model (task 2) had been speeded up on the heterogeneous platform HybriLIT by more than 200 times (from 3 days to 14 minutes) by using the cuStateVec library on GPU A100.



### hlit.jinr.ru



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