

An Introduction to  
**Quantum Annealing**

Alex Koszegi  
Solutions Architect  
[akoszegi@dwavesys.com](mailto:akoszegi@dwavesys.com)

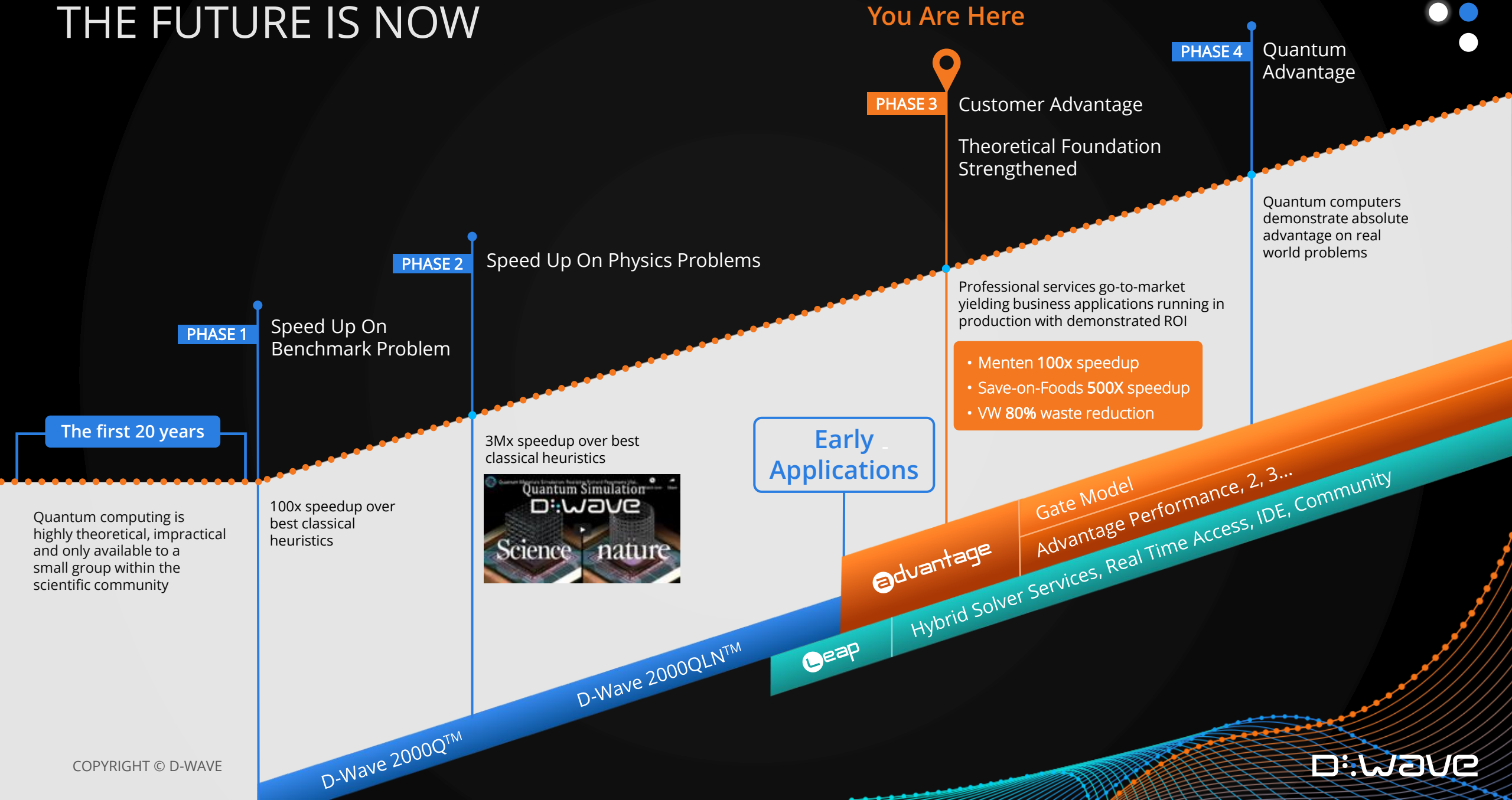
# Outline

1. D-Wave
2. Applications
3. Quantum Annealing
4. D-Wave's Implementation
5. Lab Tour
6. Q&A

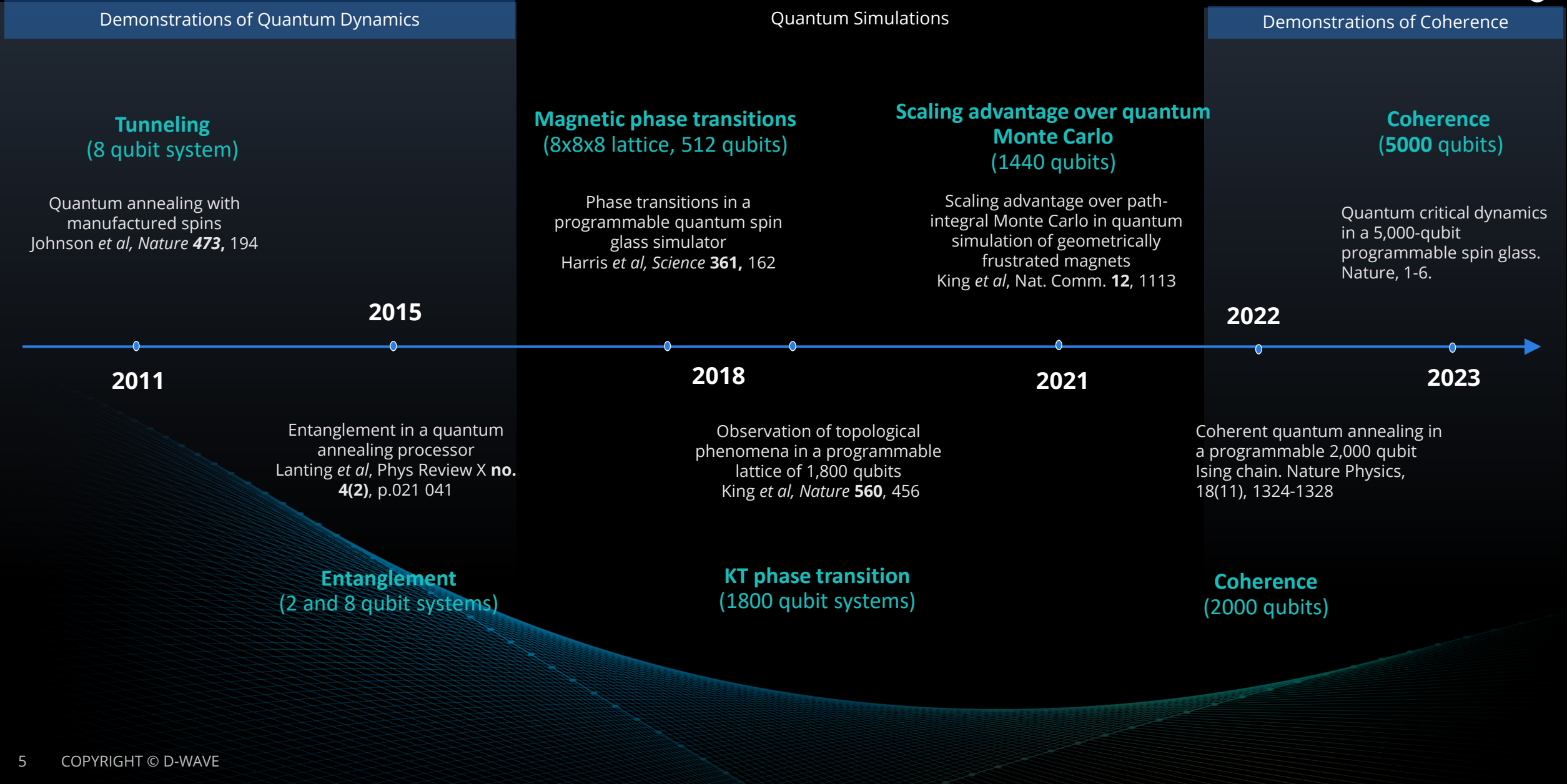


# High Level Overview: D-Wave

# THE FUTURE IS NOW



# QUANTUM TECHNOLOGY MILESTONES



# Built for Business

**D-WAVE**  **advantage**

The most connected and powerful quantum computer

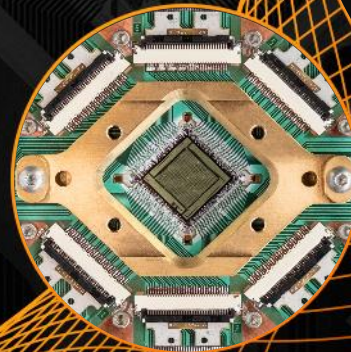
Supports hybrid applications of real-world size

- Up to 1 million variables

Annealing quantum processor design

- 5,000+ qubits

Ongoing increases in coherence & connectivity



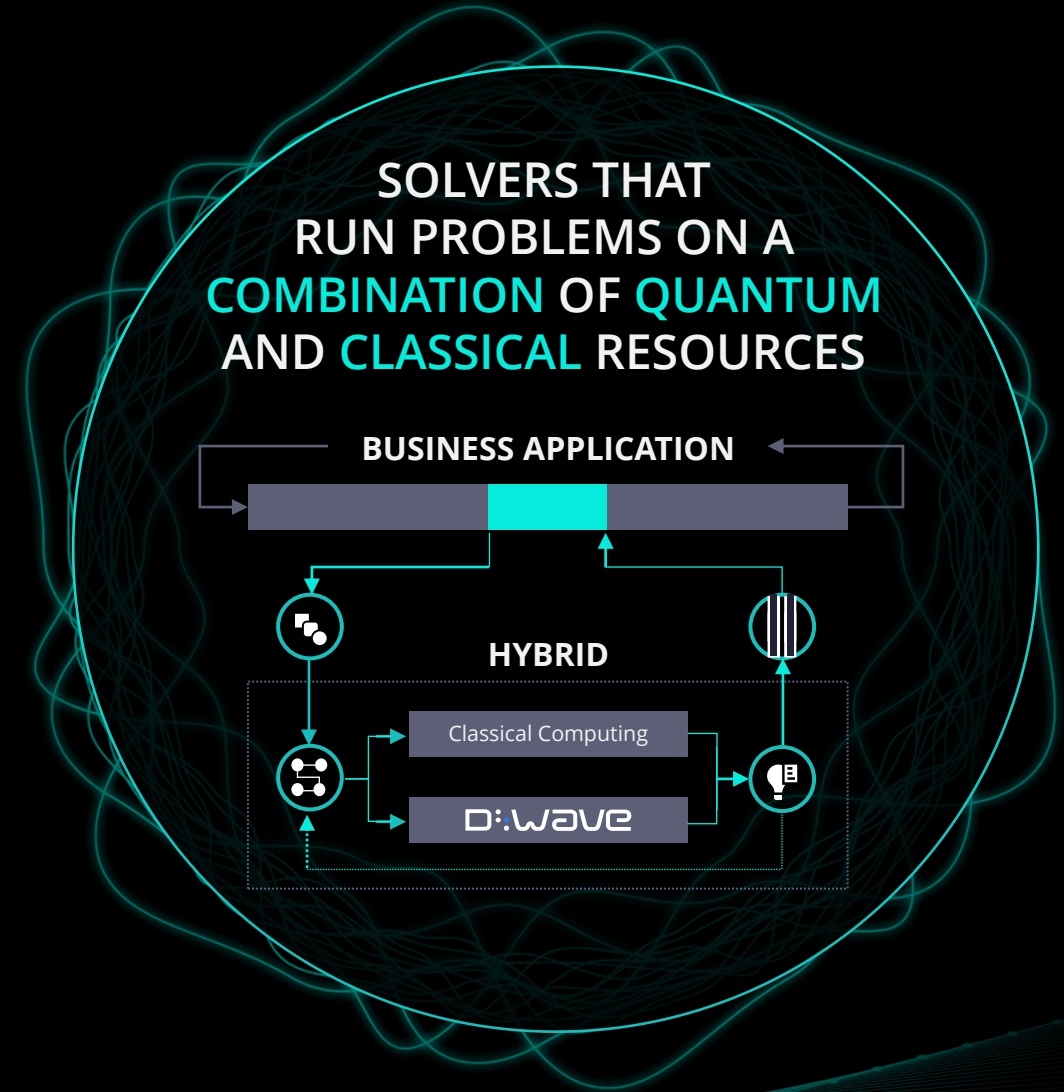
# POWERFUL HYBRID SOLVERS

## CONSTRAINED QUADRATIC MODEL SOLVER

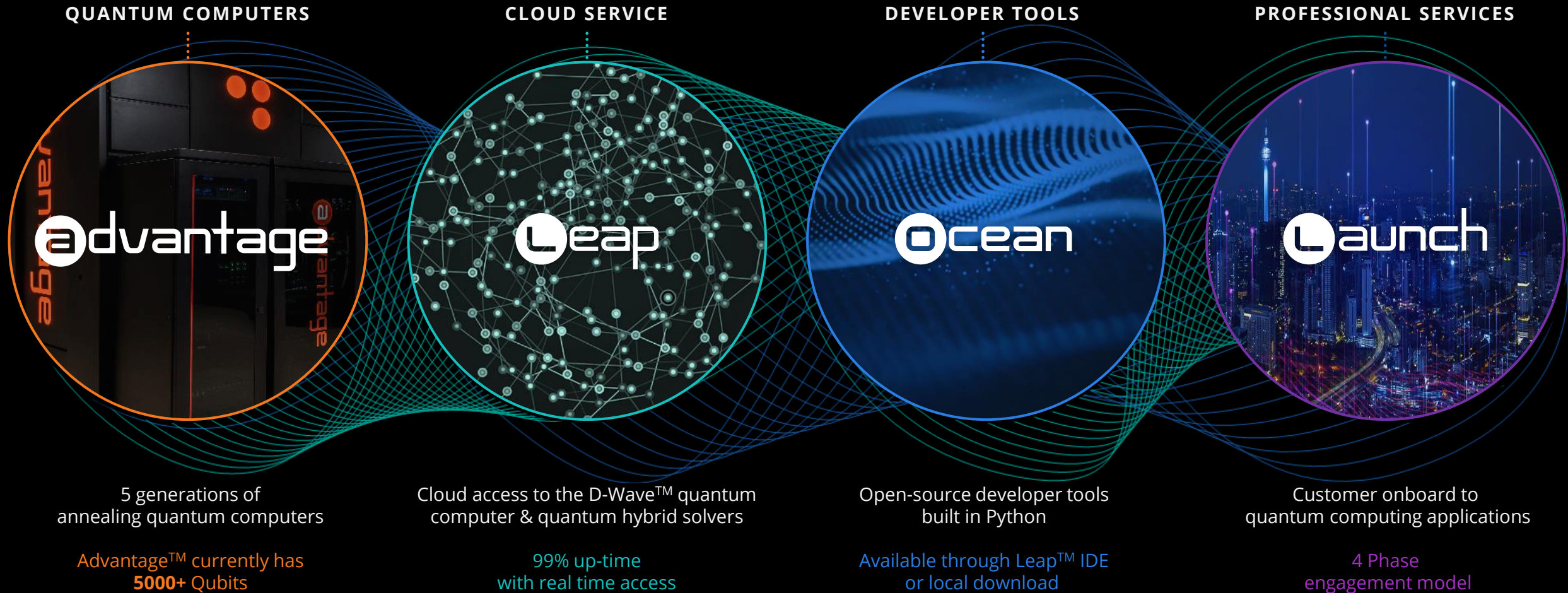
- More native representation of problem
- Unlocks larger application problems
- Inequality & equality constraints; Up to 100,000
- Binary, integer, and real/continuous variables

## BINARY QUADRATIC MODEL SOLVER

- Up to 1,000,000 variables
- Enables enterprise-scale problem solving
- Accepts problems with binary variable



# A Comprehensive, Full-Stack, Commercial Quantum Platform



# Applications

# REAL-WORLD, COMMERCIAL APPLICATIONS TODAY ACROSS KEY VERTICALS



## LOGISTICS

Shipping container logistics  
Employee scheduling  
Farm to market food delivery  
Last mile vehicle routing

## PHARMA

Protein folding  
Clinical trials  
Drug discovery

## FINANCE

Portfolio risk reduction and  
return optimization  
Marketing campaign optimization  
Fraud detection

# AUTOMATE E-COMM DRIVER SCHEDULING



## BUSINESS IMPACT:

Pattison Food Group provides next day delivery of food and health products across western Canada and the US. Schedules were manually created for over 100 stores and 500 drivers until October 2022, when its quantum-hybrid e-commerce driver auto-scheduler (QEDA) was deployed to production.

## QUANTUM-HYBRID SOLUTION:

QEDA pairs drivers with location-specific shifts so that the business needs are met (even when understaffed), driver shift and location preferences are accommodated, and schedules are consistent week to week.

## BUSINESS VALUE:

This solution meets all business requirements, reduces manual scheduling time by 80% and allows schedulers to easily adapt to last minute changes due to sickness and absence.

**“We will be the first grocery chain to actually use quantum computing to serve our business needs.”**

—Benny Wai, Manager of Analytics Development  
at Pattison Food Group

**Pattison  
Food Group**

# LOGISTICS OPTIMIZATION AT PORT OF LA



60%

MORE CARGO  
HANDLED EACH DAY  
PER CRANE

## SAVANTX

D-Wave's quantum system is used as part of the SavantX HONE optimization engine at the Port of Los Angeles. The goal is to expedite delivery of containers out of the terminal while increasing the amount of cargo that can be handled.

"With HONE and D-Wave, each huge crane handled 60% more cargo per day, while the turnaround time for trucks was reduced by 12%."

— SAVANTX TEAM



# Main Principles: Quantum Annealing

# Quantum Annealing

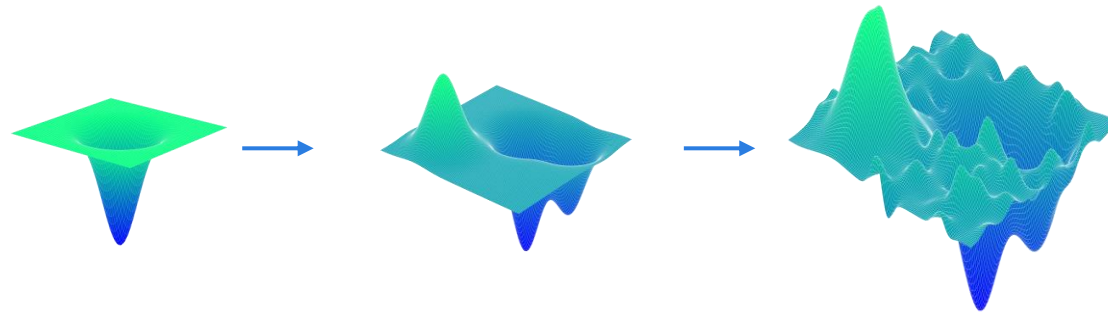


## What is quantum annealing?

- A process that uses quantum mechanical effects to find a global minimum of an objective function
- Naturally solves optimization problems

## Adiabatic Theorem

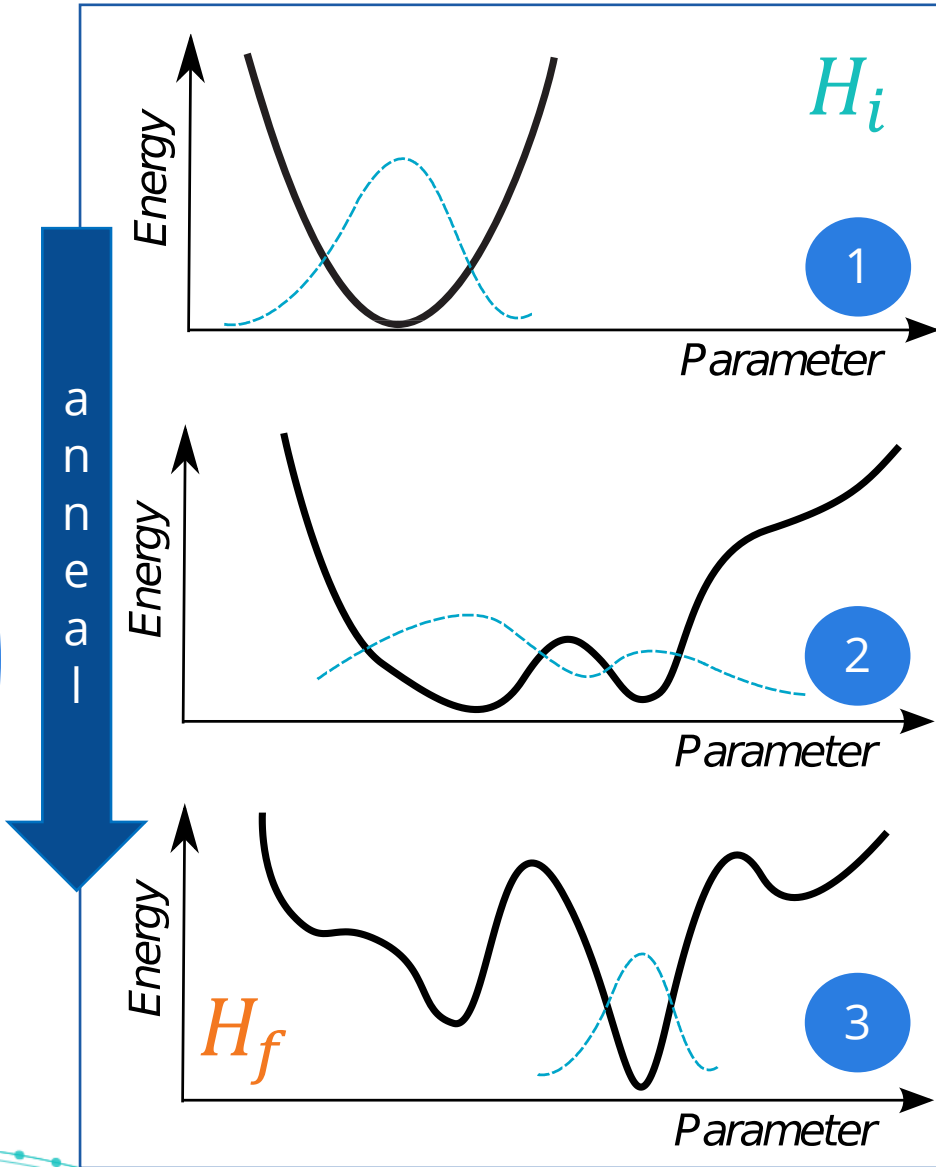
- Start a quantum system in the ground-state of a Hamiltonian (energy landscape)
- If that Hamiltonian changes slowly enough in time, the system will end up in the ground state of the final Hamiltonian



# Quantum Annealing



- 2
- Multiple shallow minima with narrow barriers that allow tunneling
  - Wavefunction delocalized across them



- 1
- Single global minimum
  - Ground state

- 3
- Many deep minima
  - Wavefunction localized around global minimum
  - The ground state is a classical spin state

# Quantum Annealing - Transverse Field Ising Hamiltonian



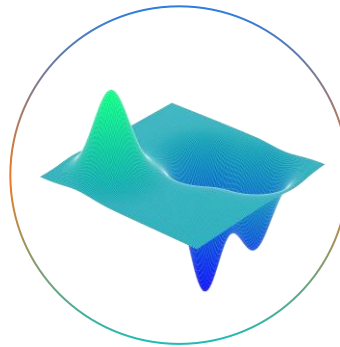
$$H_{ising} = \underbrace{\frac{-A(s)}{2} \left( \sum_i \sigma_x^{(i)} \right)}_{\text{Transverse Field Hamiltonian}} + \underbrace{\frac{B(s)}{2} \left( \sum_i h_i \sigma_z^{(i)} + \sum_{i>j} J_{i,j} \sigma_z^{(i)} \sigma_z^{(j)} \right)}_{\text{Problem Hamiltonian}}$$

Transverse Field Hamiltonian

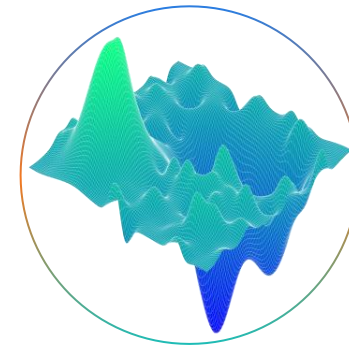
Problem Hamiltonian



Transverse Field Hamiltonian



Transverse Field + Problem Hamiltonian

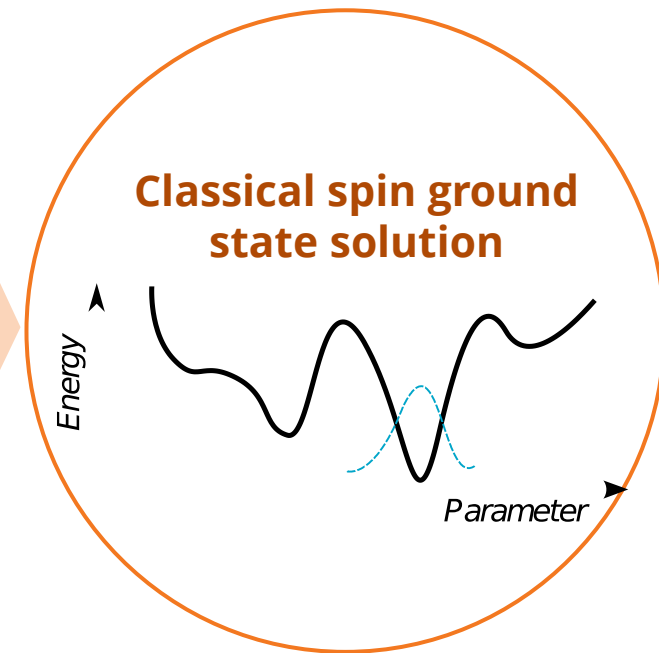
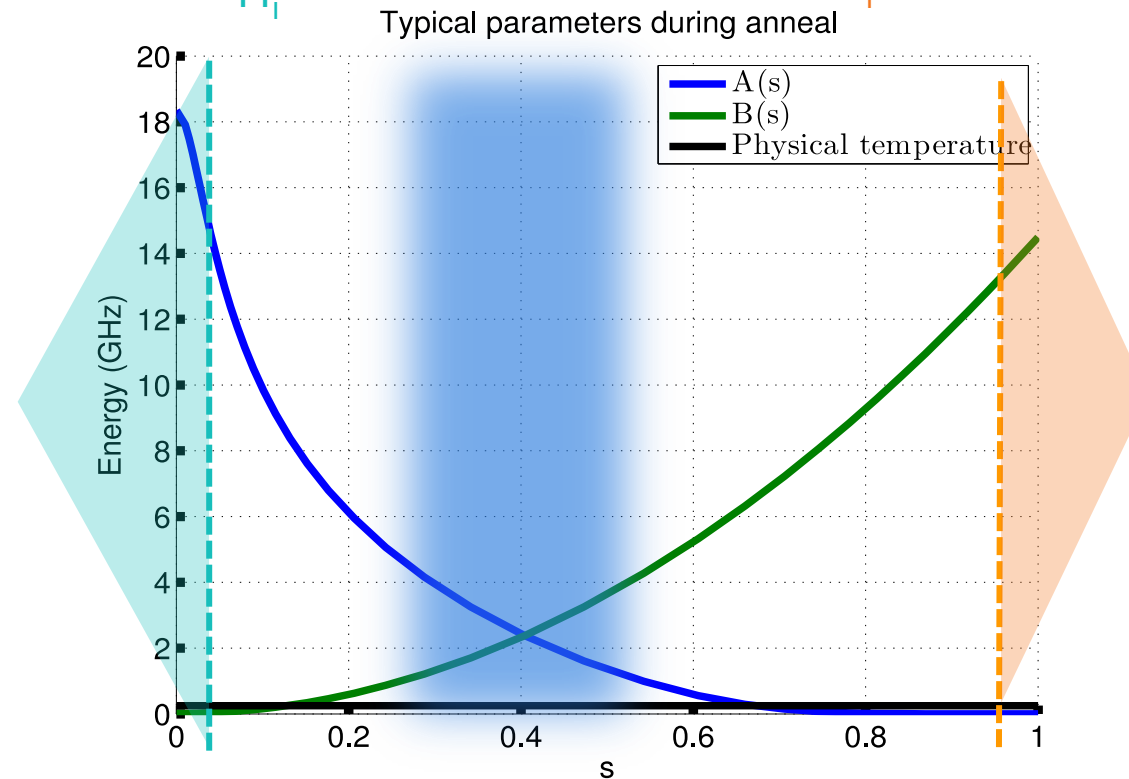
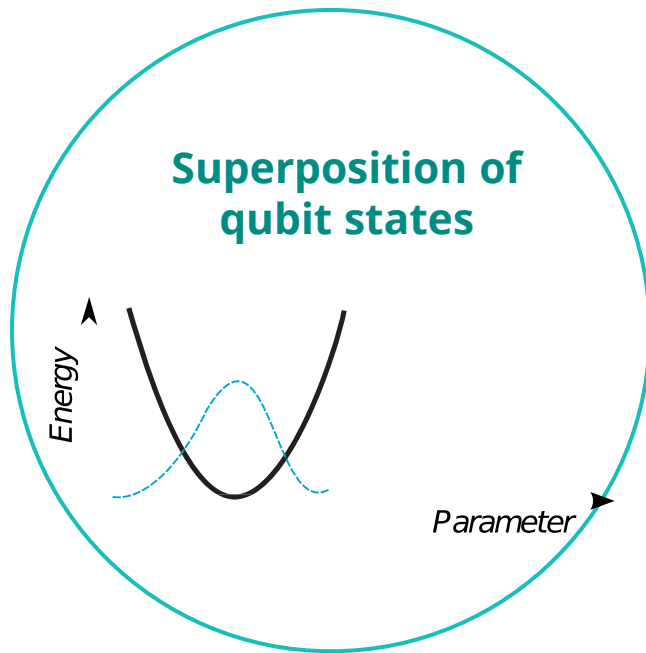


Problem Hamiltonian

# Quantum Annealing - Transverse Field Ising Hamiltonian



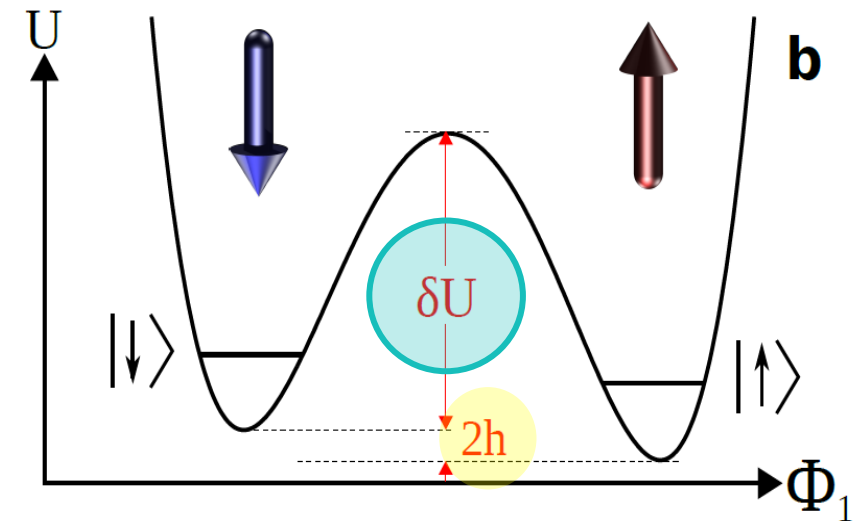
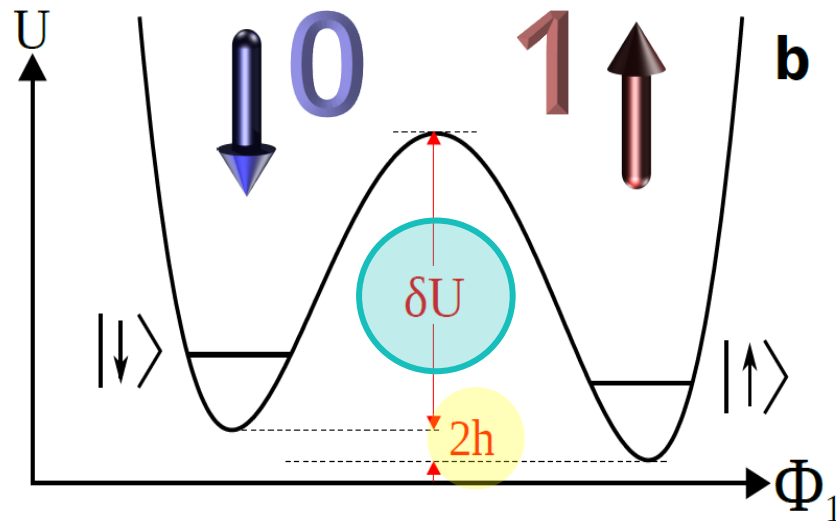
$$H(s) = \underbrace{-A(s) \left[ \sum_i \sigma_x^{(i)} \right]}_{H_i} + \underbrace{B(s) \left[ - \sum_i h_i \sigma_z^{(i)} + \sum_{i,j>i} J_{ij} \sigma_z^{(i)} \sigma_z^{(j)} \right]}_{H_f}$$



# Quantum Annealing - Transverse Field Ising Hamiltonian



$$H(s) = -A(s) \left[ \sum_i \sigma_x^{(i)} \right] + B(s) \left[ - \sum_i h_i \sigma_z^{(i)} + \sum_{i,j>i} J_{ij} \sigma_z^{(i)} \sigma_z^{(j)} \right]$$



$h_i$  and  $J_{ij}$  are specified with **programmable on-chip** control circuitry



# Main Principles: D-Wave's Implementation

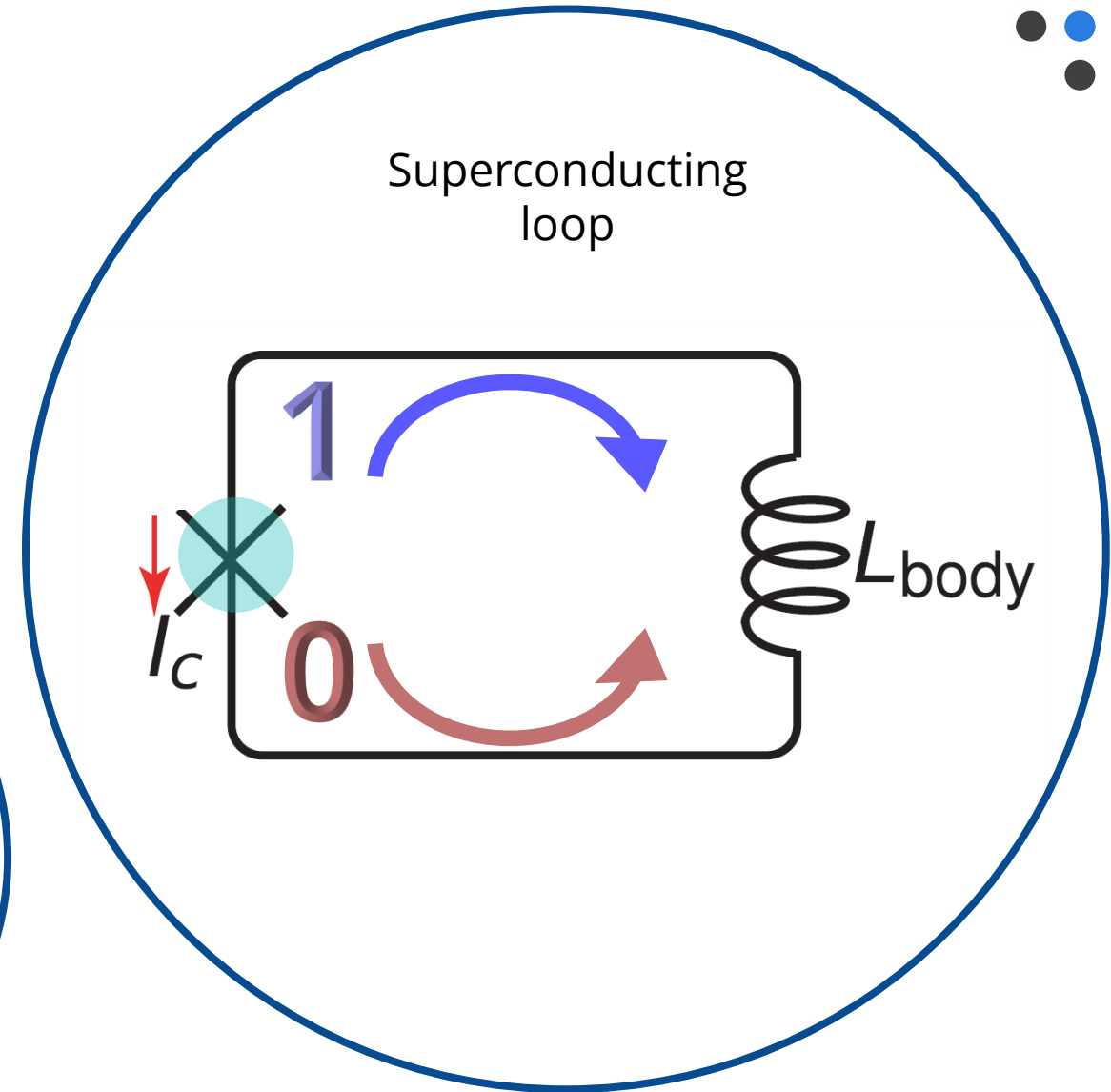
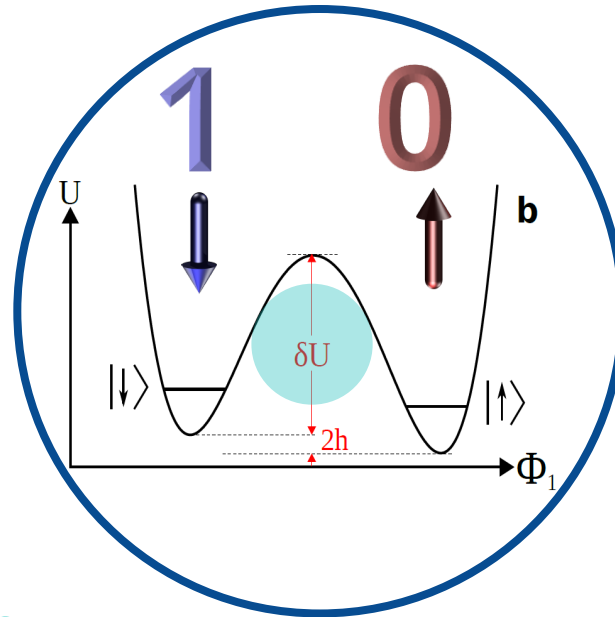
# The Qubit – RF SQUID

## Josephson Junction

creates barrier in potential

## Critical current $I_c$

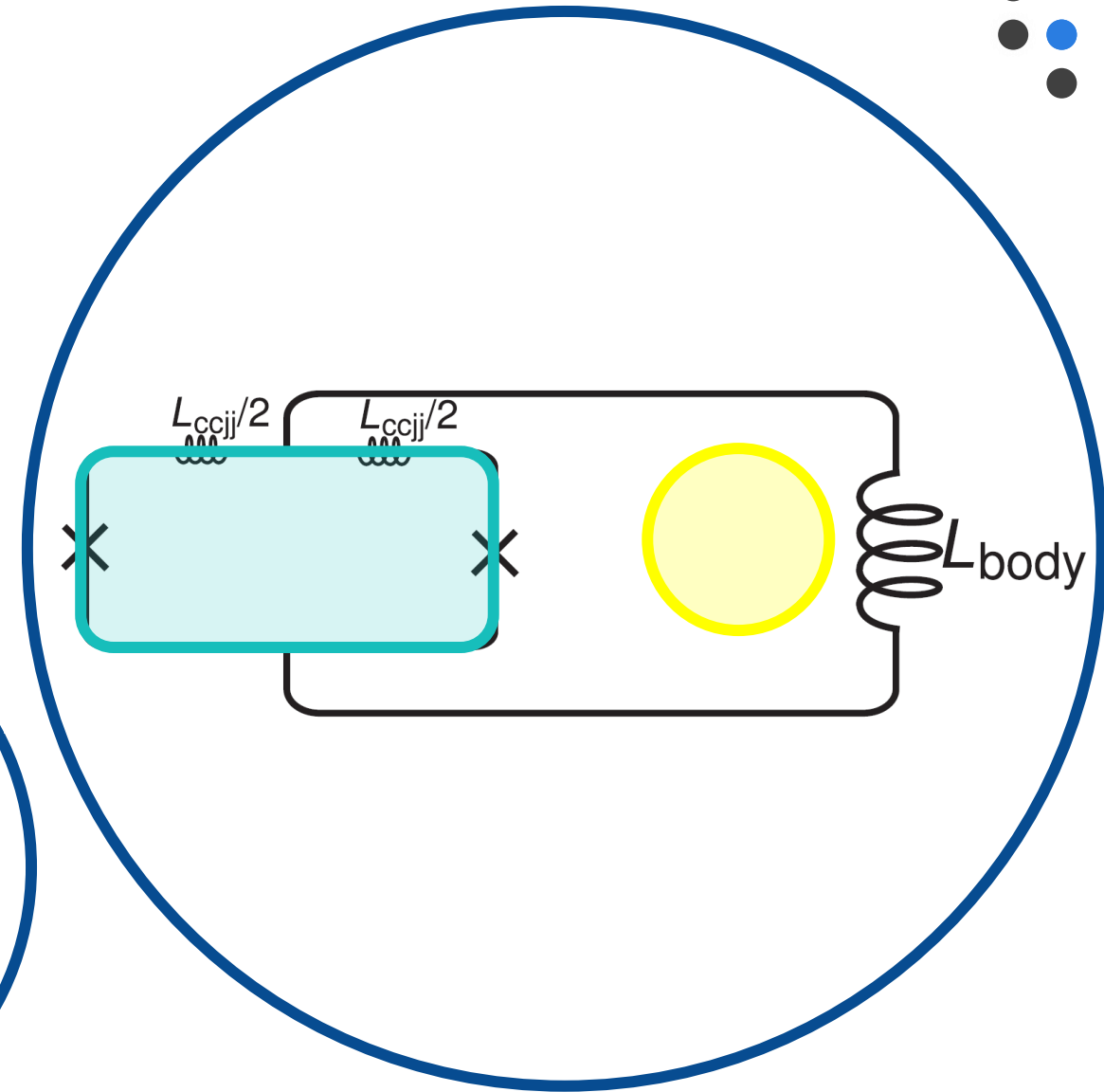
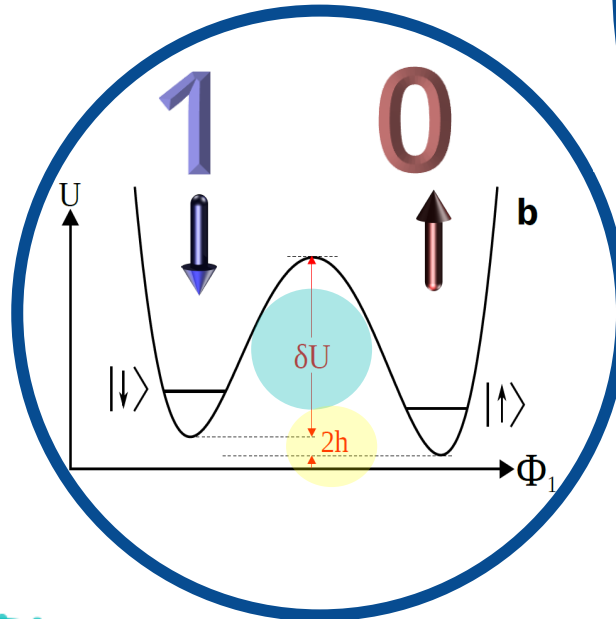
determines the barrier height



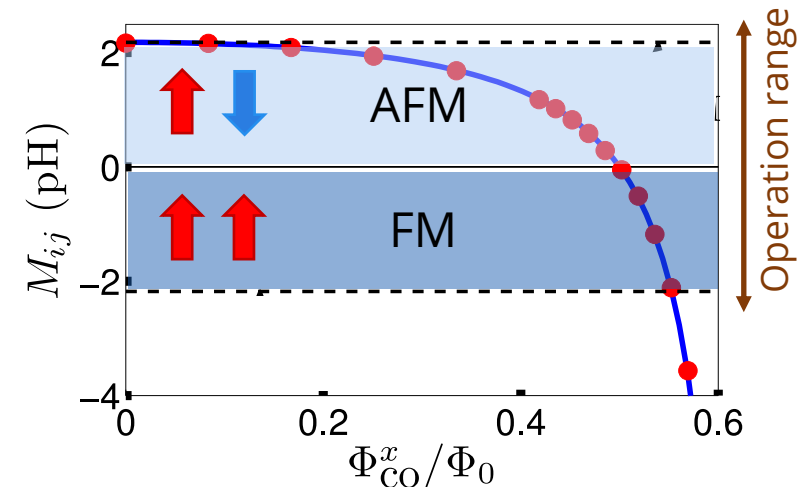
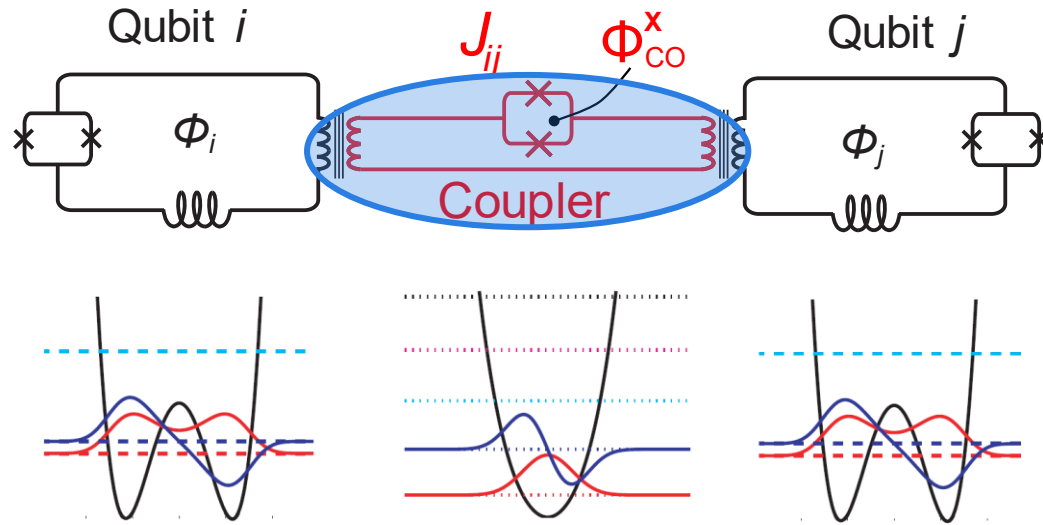
# The Qubit – RF SQUID

$$H(s) = -A(s) \left[ \sum_i \sigma_x^{(i)} \right] + B(s) \left[ -\sum_i h_i \sigma_z^{(i)} + \sum_{i,j>i} J_{ij} \sigma_z^{(i)} \sigma_z^{(j)} \right]$$

Compound Josephson Junction (CJJ) creates a **tunable** barrier



# The Qubit – Coupling Qubits

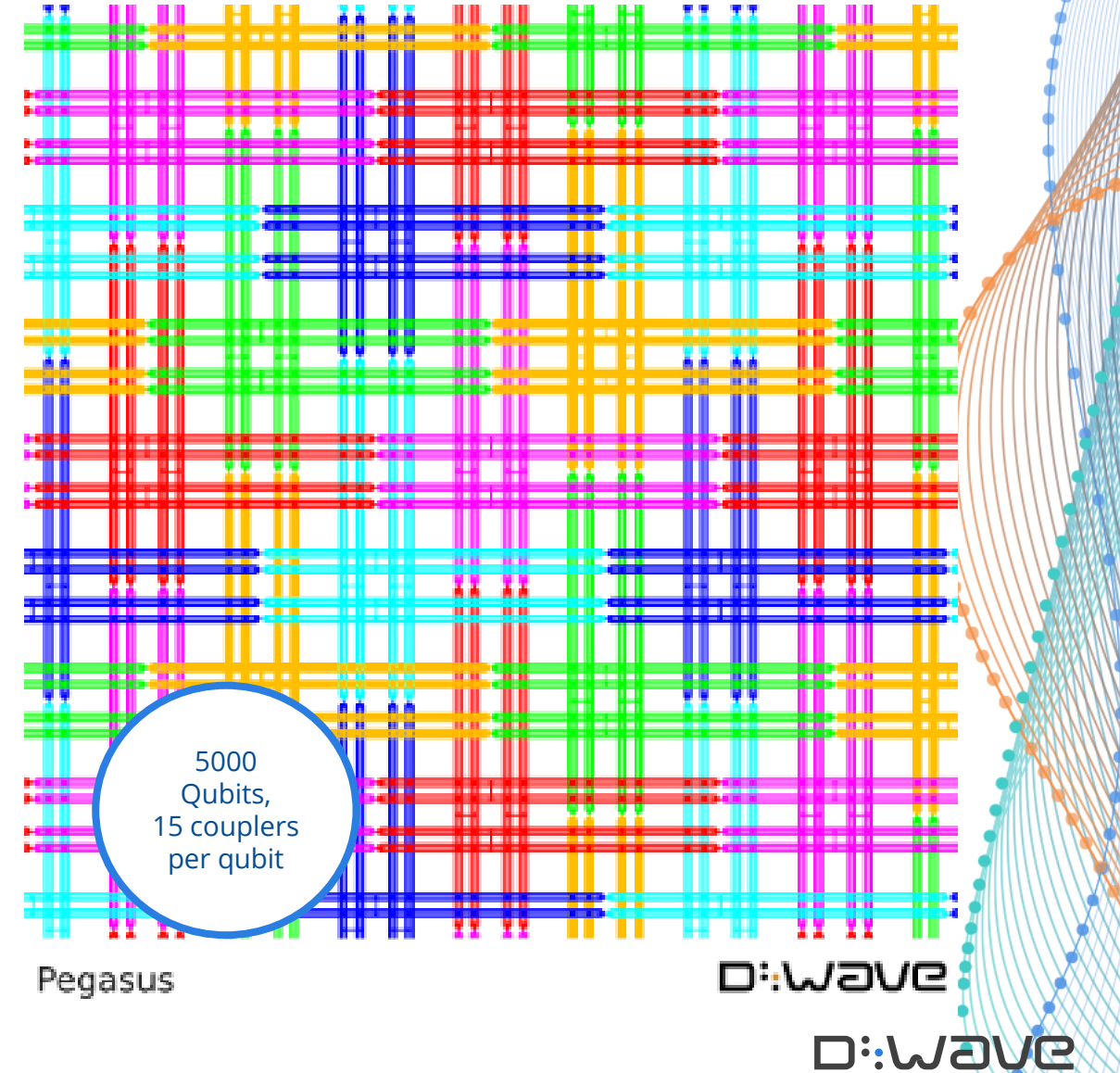
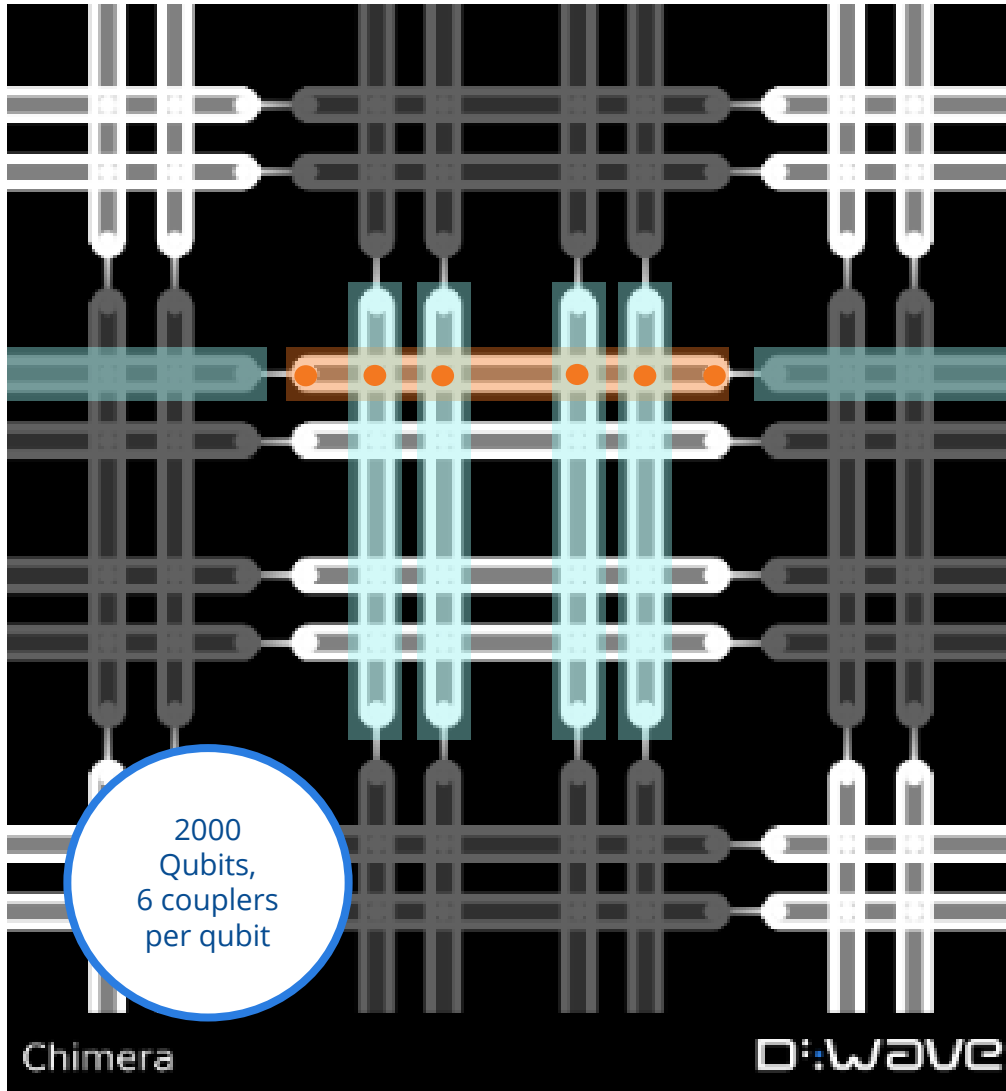


A mono-stable two-junction RF SQUID provides a tunable mutual inductance

$$M_{ij} = \frac{J_{ij}}{|I_p(t)|^2}$$

$$H(s) = -A(s) \left[ \sum_i \sigma_x^{(i)} \right] + B(s) \left[ -\sum_i h_i \sigma_z^{(i)} + \sum_{i,j>i} J_{ij} \sigma_z^{(i)} \sigma_z^{(j)} \right]$$

# The Chip - Processor Layout



# The Chip - Processor Layout

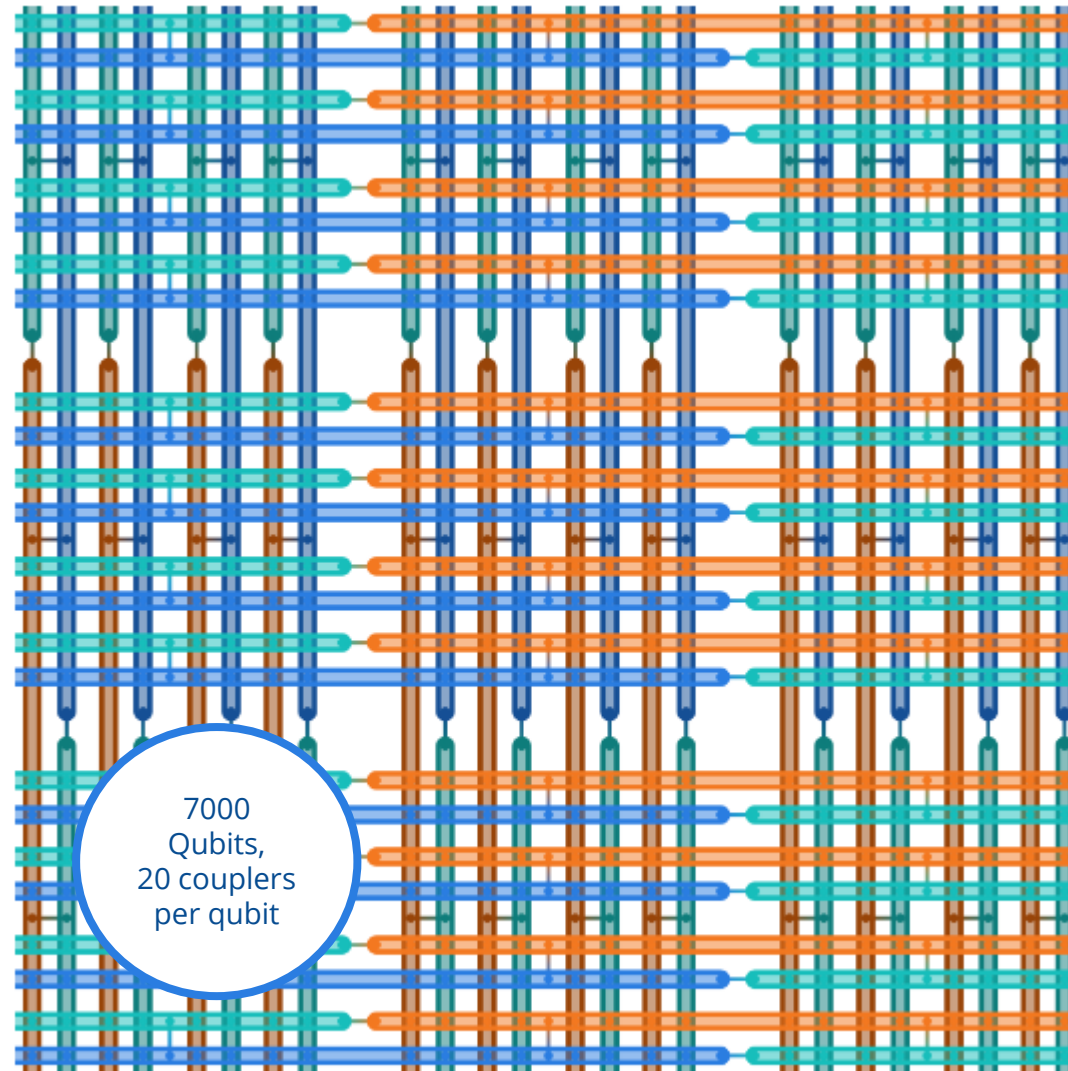


Fig. 25 Zephyr unit cells: for the center unit cell, one group of eight half qubits are shown in orange, another in blue.

# Lab Tour!

## Lab Tour:

Video 1: <https://youtu.be/zDotDiK2UuY?t=1>

Video 2: <https://youtu.be/AGByZoYUIU0>

# Differences Between Annealing and Gate



	Quantum Annealing	Gate
The Foundation	<ul style="list-style-type: none"><li>Physical implementation of a class of quantum algorithms</li><li>Continuously interpolates between uniform <math>\sigma^x</math> and user-defined <math>\sigma^z</math></li></ul>	<ul style="list-style-type: none"><li>Can perform different algorithms, some of which have shown a theoretical quadratic or exponential speed up over classical algorithms for certain problems</li><li>Full access to operations on the Bloch sphere</li></ul>
Hardware Scalability	<ul style="list-style-type: none"><li>Easier to fabricate and calibrate larger scale processors</li><li>Difficult to achieve full connectivity and reduce intrinsic noise sources (flux and charge noise)</li><li>Naturally fault tolerant</li></ul>	<ul style="list-style-type: none"><li>Difficult to fabricate larger scale processors due to need for increased coherence times</li><li>Difficult to achieve fully error corrected systems</li></ul>
Use Cases	<ul style="list-style-type: none"><li>Optimization problems<ul style="list-style-type: none"><li>Scheduling, portfolio optimization, logistics</li></ul></li><li>Can simulate certain quantum systems (Quantum spin ice, Shastry-Sutherland Ising Model, KT phase transition simulation)</li></ul>	<ul style="list-style-type: none"><li>General applications<ul style="list-style-type: none"><li>Material science, quantum chemistry, drug discovery</li></ul></li><li>Can simulate quantum systems</li></ul>

# Quantum Annealing vs QAOA



	Quantum Annealing	Quantum Approximate Optimization Algorithm (QAOA)
General	Continuously changes from a driver to the problem Hamiltonian.	Attempts to approximate quantum annealing by mixing a driver and problem Hamiltonian over discrete iterations.
Computation Time	Typical operation times below ~100 ms for $N \leq 5000$ .	Experiments show operation times of minutes for $N \leq 50$ due to NP-hard algorithmic parameter tuning.
Solution Quality	Consistently shows higher probability of sampling the ground state (>90%) in studies comparing against QAOA.	Empirical tests suggest low probability (0.01% – 80%) of finding the ground state. Highly dependent on the input and hardware.

# Summary

# Summary

D-Wave is currently the **only quantum annealing provider**.

Our current generation, Advantage, has **over 5000 qubits**. Advantage2 will have **over 7000 qubits**.

Since 2011 we have published **demonstrations of quantum effects** in our processors and **simulations that show up to a 3 million times speed up** compared to classical techniques.

Quantum annealing will likely always be the **best quantum technology for optimization problems**.

D-Wave supports quantum-hybrid **production applications today**.

# Additional Resources

Sign up for a free Leap account to access code examples and run problems today:  
<https://cloud.dwavesys.com/leap/signup/>

Learn more about our applications:  
<https://dwavequantum.com/learn/featured-applications/>

Learn more about our technology:  
<https://dwavequantum.com/learn/resource-library/>

## QA vs QAOA

- Why D-Wave is Bullish on Quantum Annealing – Catherine McGeoch  
[https://youtu.be/lAeZ\\_hs9OBw](https://youtu.be/lAeZ_hs9OBw)
- Lubinski et al., Optimization Applications as Quantum Performance Benchmarks (2023)  
<https://arxiv.org/abs/2302.02278>
- Lotshaw et al. Scaling Quantum Approximate Optimization on Near-term Hardware (2022)  
<https://arxiv.org/abs/2201.022477>