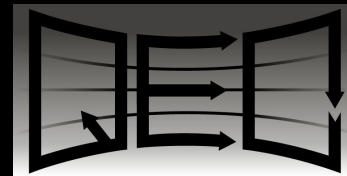


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Exponentially-Biased Ground-State Sampling of Quantum Annealing Machines with Transverse-Field Driving Hamiltonians

Dr. Salvatore Mandrà

What is fair sampling?

Definition (fair sampling):

- The ability of an algorithm to find all solutions of a degenerate problem with equal probability when run in **repetition mode**

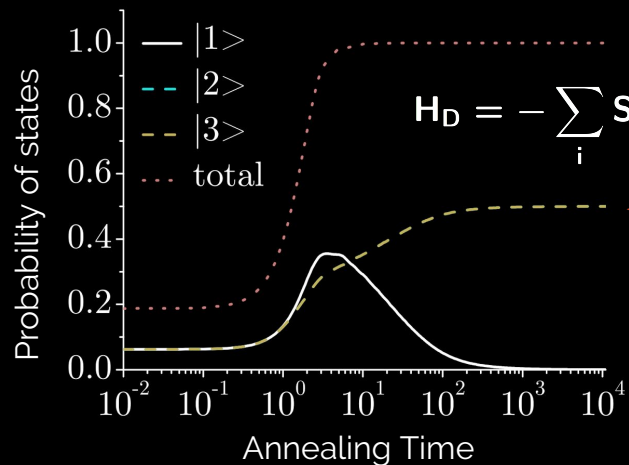
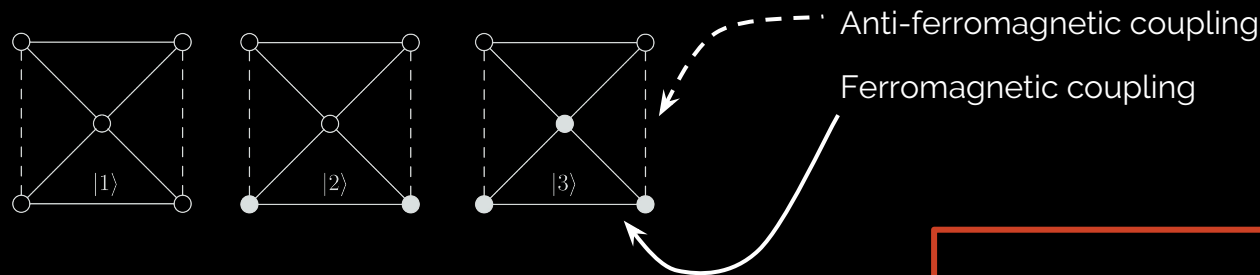
Why is it important?

- In some contexts (SAT-Filter, #SAT, machine learning, ...) finding a **good variety** of solutions is more important than finding a single solution quickly

Optimize benchmarking:

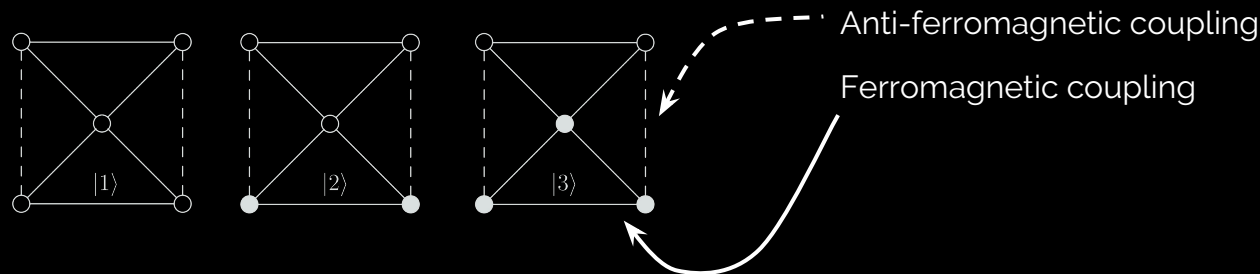
- Standard test: Find the ground-state energy **fast and reliably**
- Stringent test: Find **all minimizing configurations** equiprobably

Previous studies on transverse field QA [1]

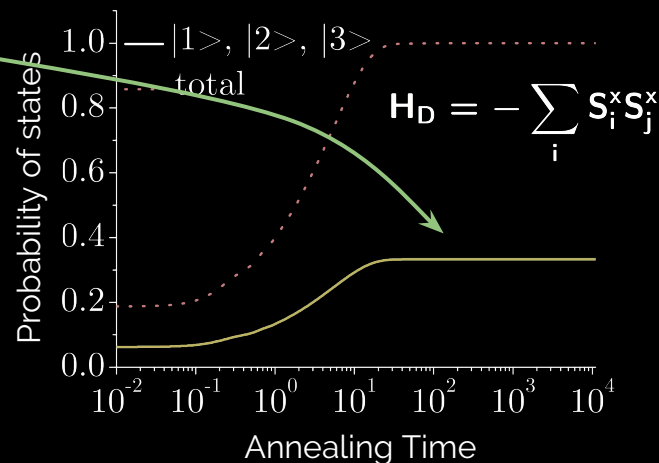
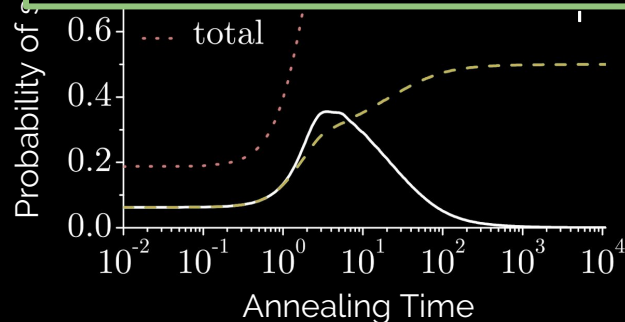


**Transverse field QA
is biased ...**

Previous studies on **transverse field QA** [1]

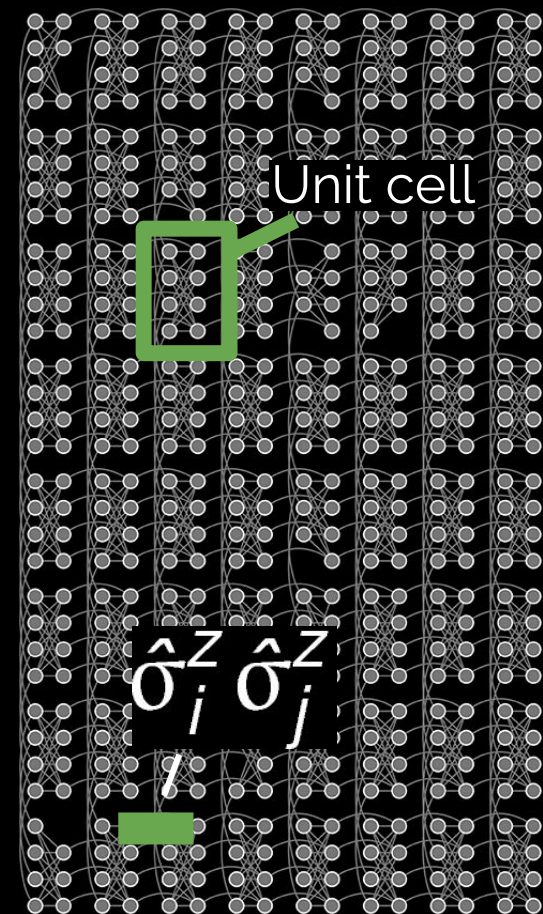


Non-stoquastic H_D mitigates the problem!



[1] Y. Matsuda, H. Nishimori & H. G Katzgraber, "Ground-state statistics from annealing algorithms: quantum versus classical approaches.", New Journal of Physics, 11(7), 073021 (2009)

The D-Wave 2X quantum annealer



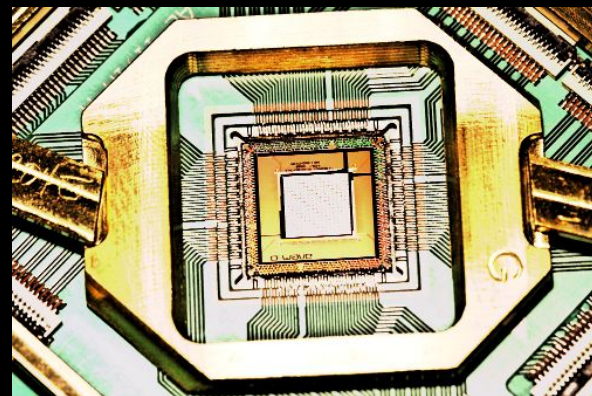
— H_p

$$H_D = - \sum_i \hat{\sigma}_i^x$$

- Unavoidable **noise**
- Non-zero **temperature**

~1000 working qubits

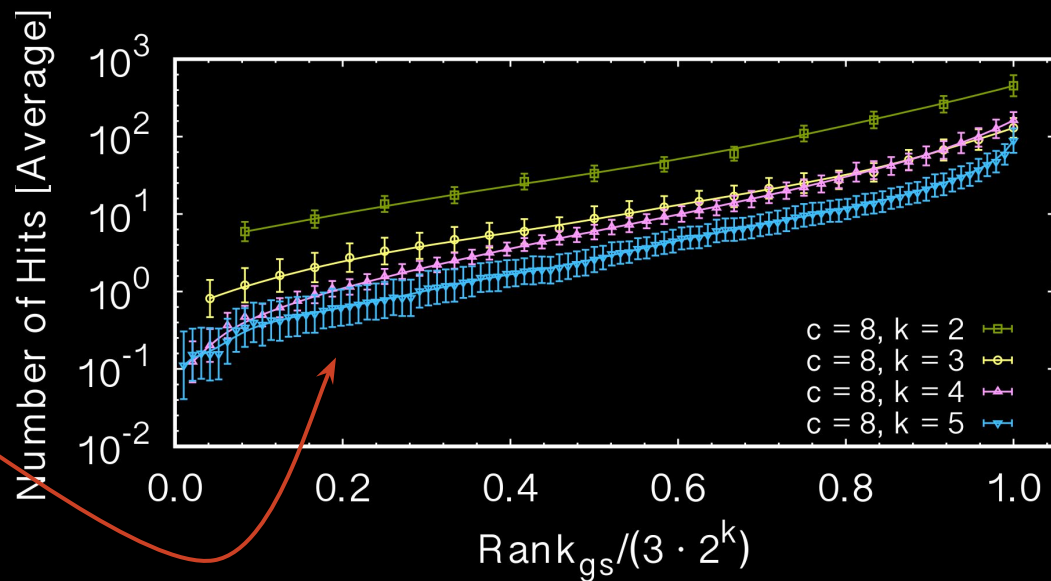
Superconducting qubit chip



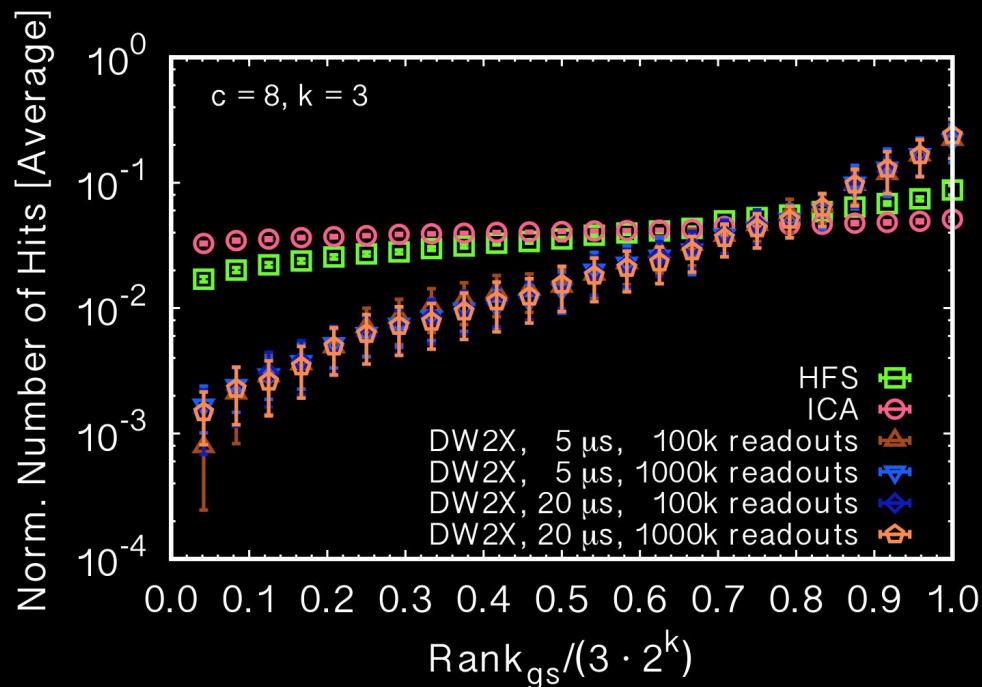
Experimental analysis using DW2X device [1]

- Random couplings from **Sidon set** ($J_{ij} = \pm 5, \pm 6, \pm 7$ on Chimera of $c \times c$ unit cells)
- Limit the study to instances with **well controlled degeneracy** ($\#_{gs} = 3 \cdot 2^k$)
- No **trivial** degeneracy
- 100 gauges x {10k, 100k} readouts
- $T_{ann} = 5\mu, 20\mu, 200\mu$

**DW2X is
exponentially biased!**

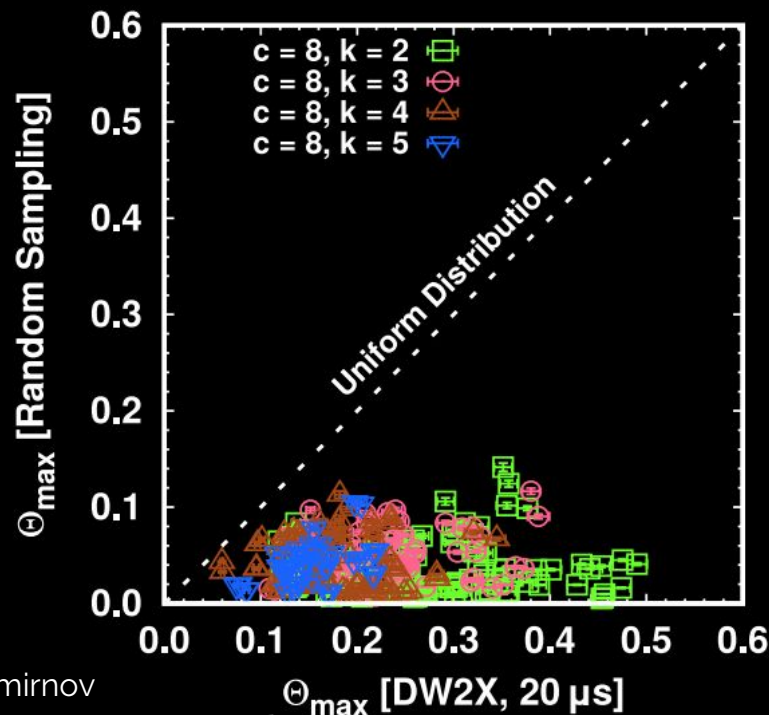


Classical algorithms sample more homogeneously

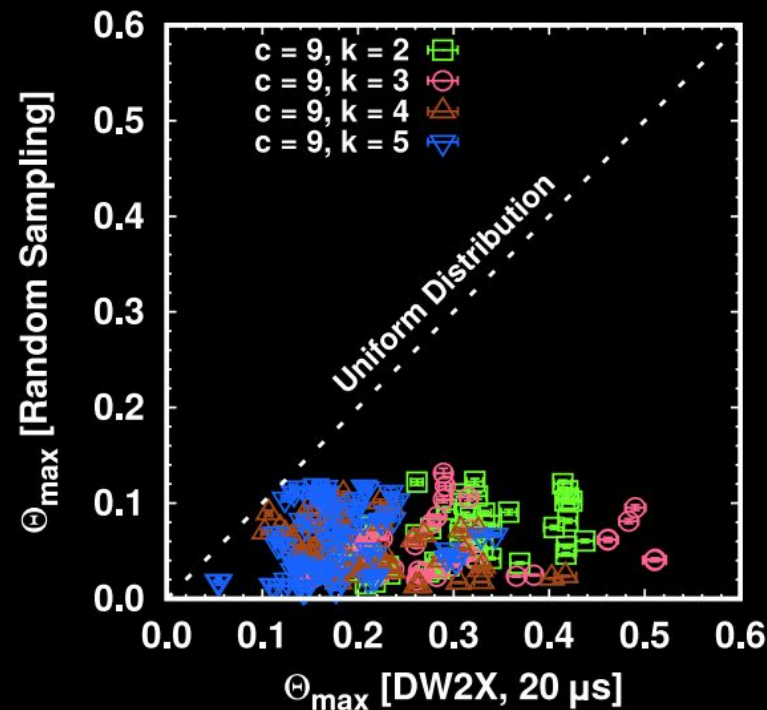


- [1] **S. Mandrà**, Z. Zhu & H. G. Katzgraber, "Exponentially-Biased Ground-State Sampling of Quantum Annealing Machines with Transverse-Field Driving Hamiltonians", arXiv:1606.07146
- [2] F. Hamze & N. de Freitas, Proceedings (2004), A. Selby, arXiv (2014)
- [3] Z. Zhu, A. J. Ochoa & H. G. Katzgraber, PRL (2015)

Experimental analysis using DW2X device [1]



Kolmogorov-Smirnov
Test

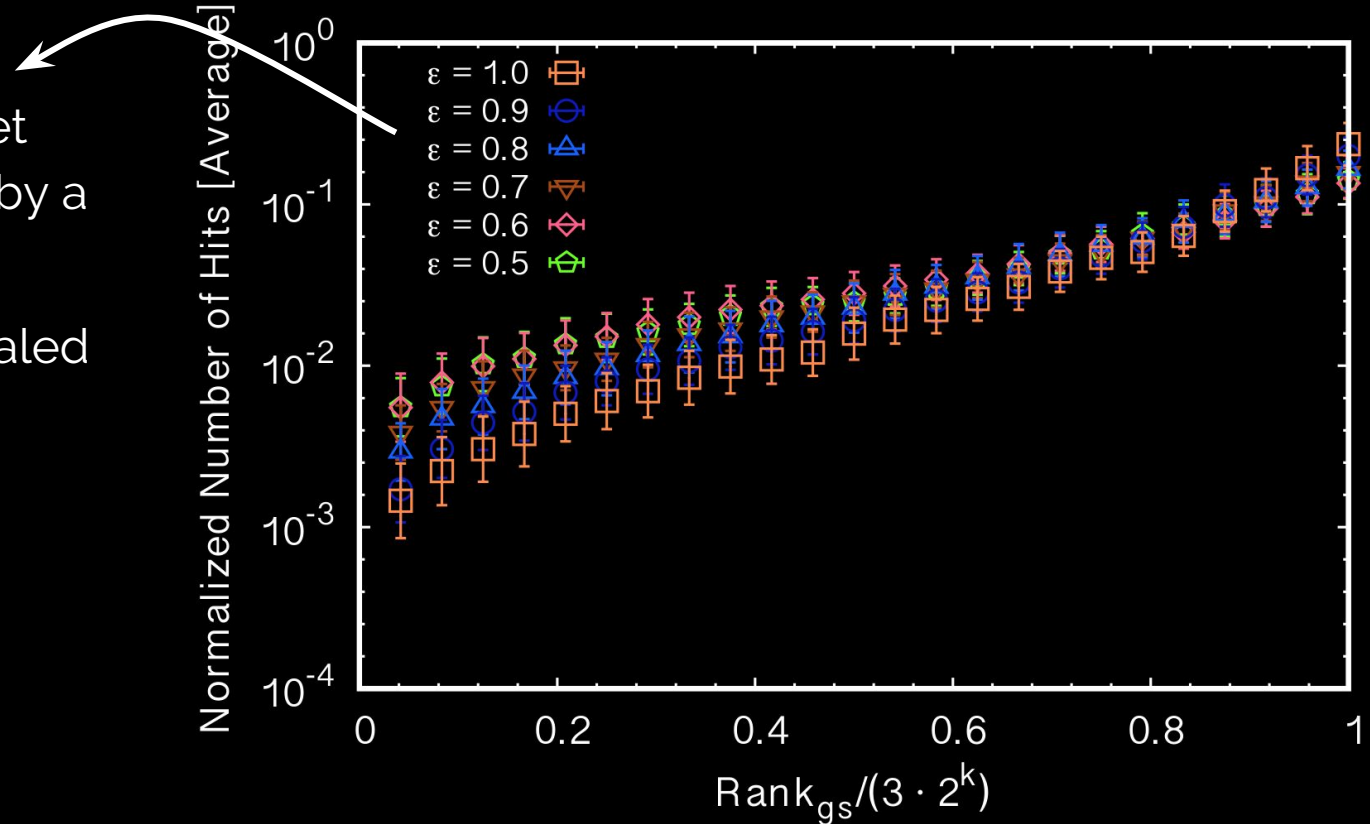


Could the **bias** be a consequence of
the **intrinsic noise** of the DW2x?

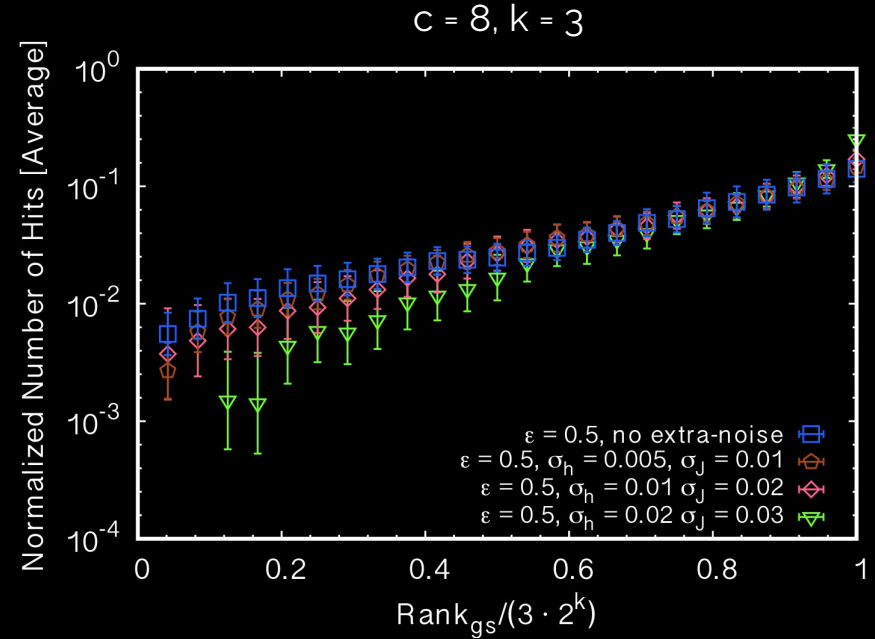
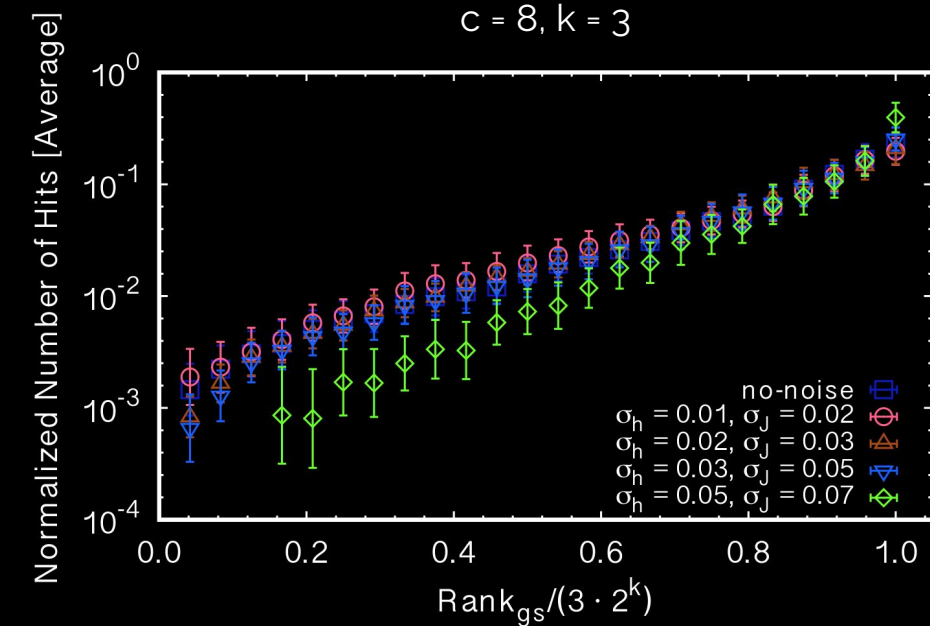
No.

The bias is **unchanged** by rescaling the energy

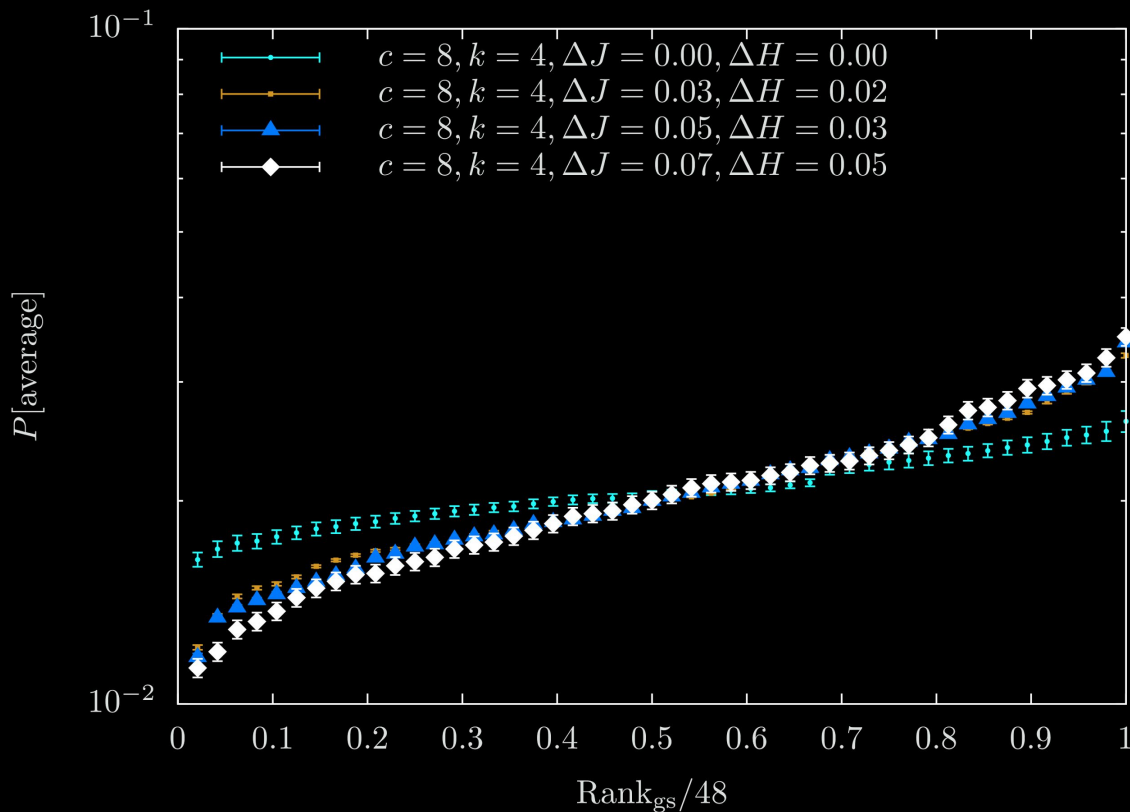
- Energy of the target problem rescaled by a factor ε
- Intrinsic noise rescaled by a factor $1/\varepsilon$



Adding extra noise does not change the bias

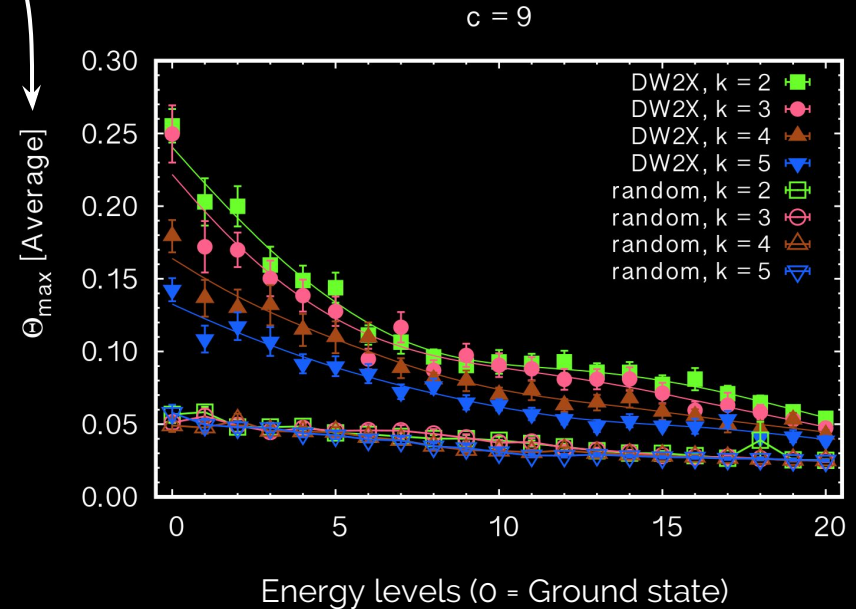
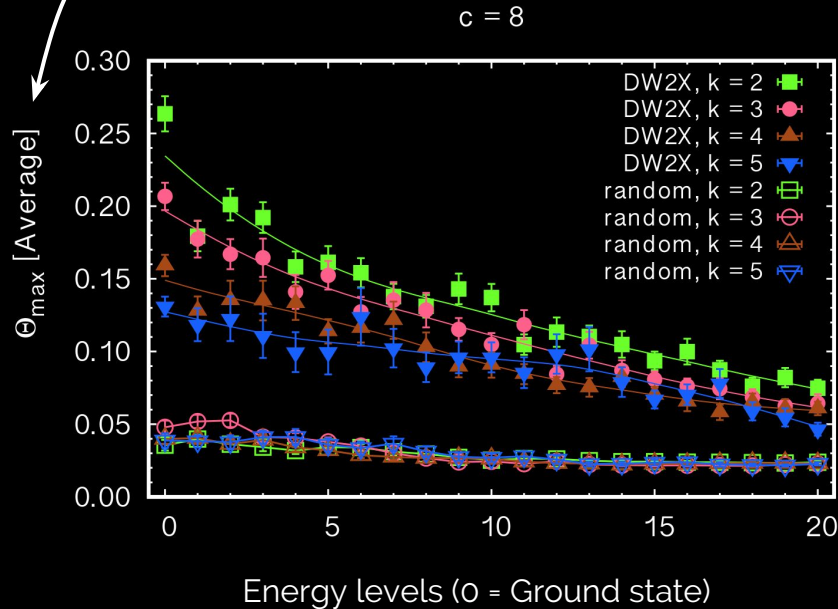


Classical algorithms are marginally affected by the noise



The bias persists up to the 20th excited state!

Different of the sampling respect to the flat distribution (**larger is worse**)



Implications & Future directions

The bias can limit the use of QA for sampling

- Applications like SAT-Filter and machine learning may not be suitable for QA without mitigating the sampling problem

How to mitigate the sampling problem?

- Explore different driver Hamiltonians (e.g. non-stoquastic)

How to understand the bias problem better?

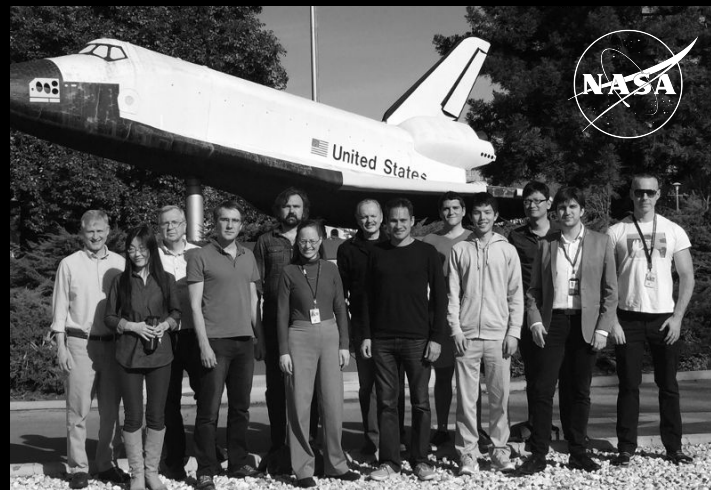
- Theoretical understanding of the role of the driver Hamiltonian in sampling
- Theoretical exploration of the implication of many-body localization



Zheng Zhu
Texas A&M



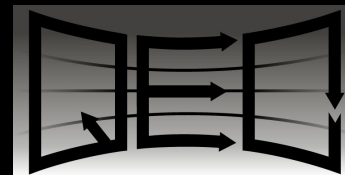
Helmut G.
Katzgraber
Texas A&M



NASA QUAIL



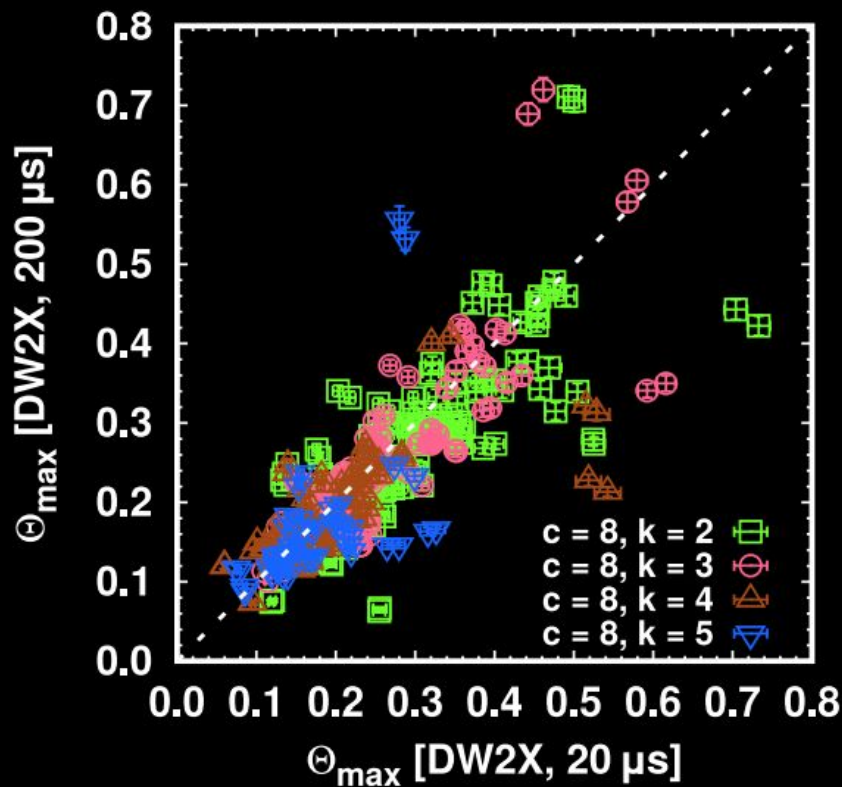
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Thanks for the attention!

Experimental analysis using DW2X device [1]



[1] **S. Mandrà**, Z. Zhu & H. G. Katzgraber, "Exponentially-Biased Ground-State Sampling of Quantum Annealing Machines with Transverse-Field Driving Hamiltonians", arXiv:1606.07146

Adiabatic Quantum Optimization (AQO)

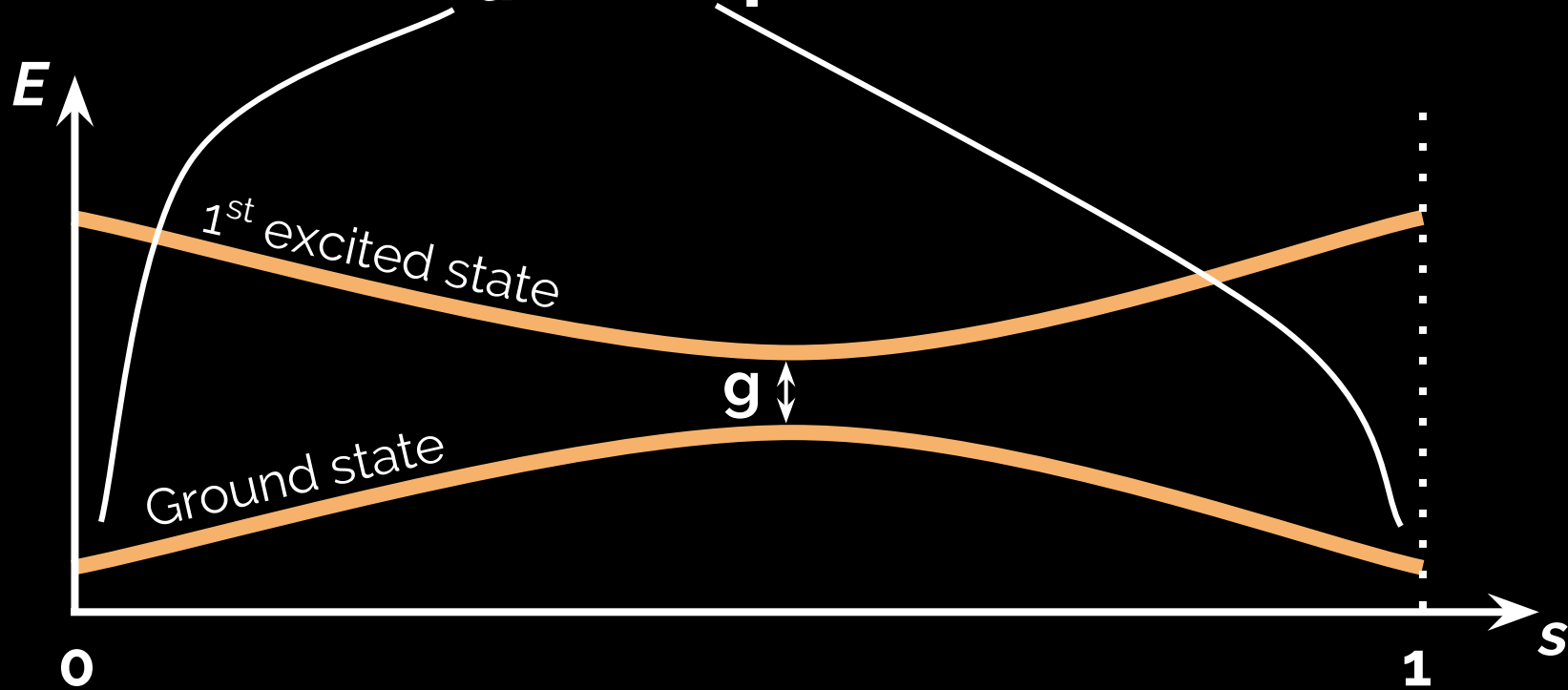
$$\mathbf{H} = (\mathbf{1} - \mathbf{s})\mathbf{H}_d + \mathbf{s}\mathbf{H}_p$$

Initial “driver”
Hamiltonian

Target Problem

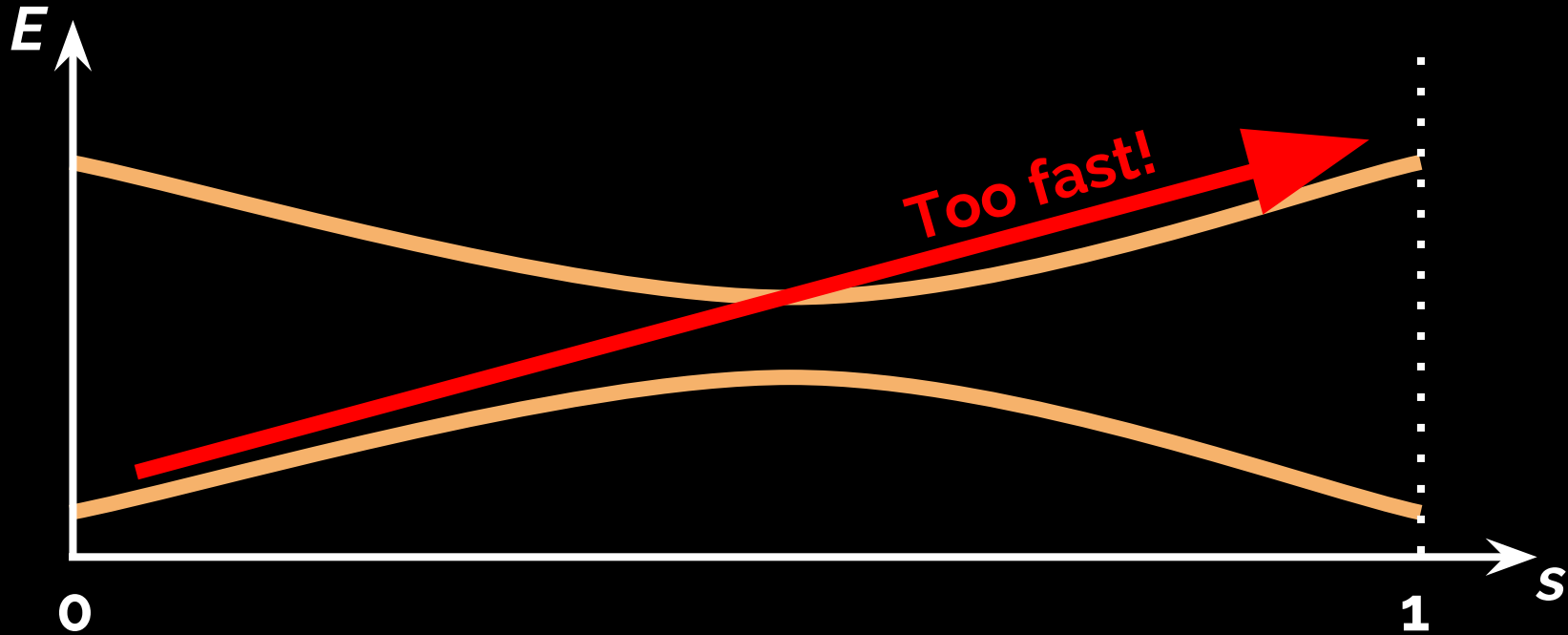
Adiabatic Quantum Optimization (AQO)

$$H = (1 - s)H_d + sH_p$$



Adiabatic Quantum Optimization (AQO)

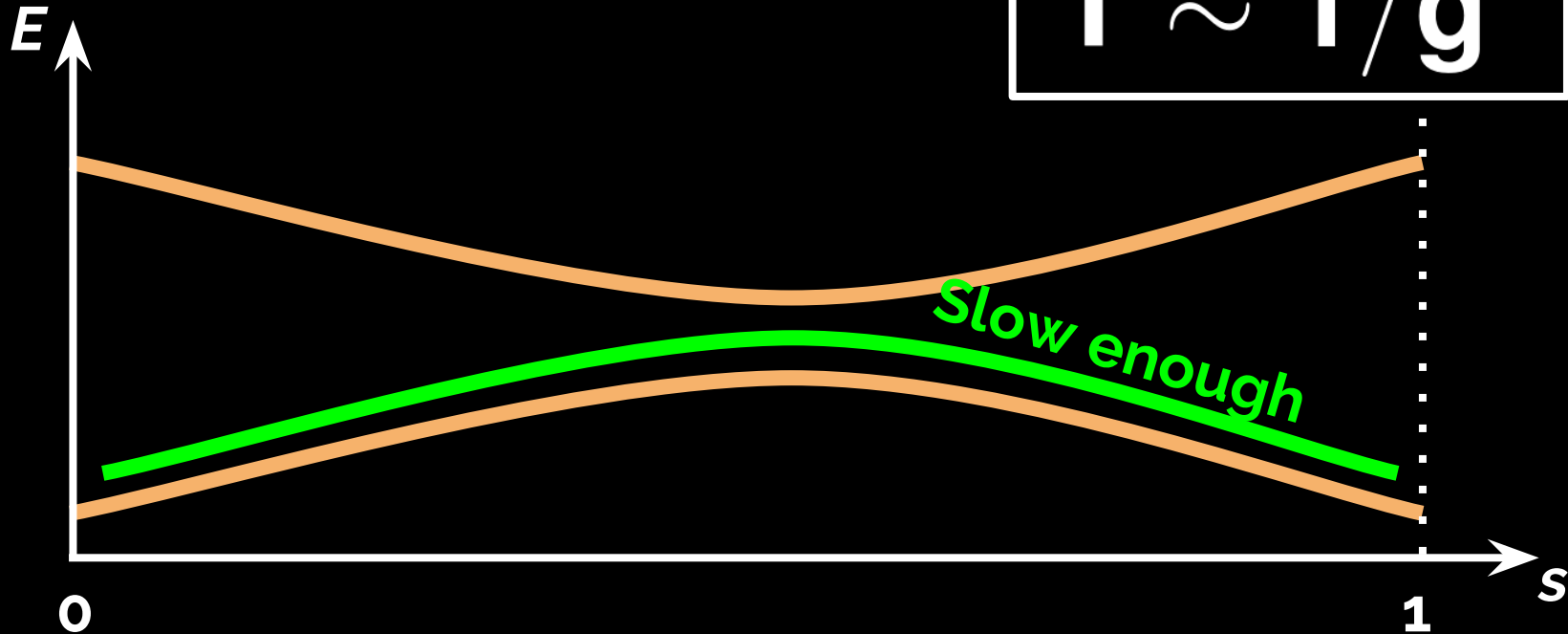
$$H = (1 - s)H_d + sH_p$$



Adiabatic Quantum Optimization (AQO)

$$H = (1 - s)H_d + sH_p$$

$$T \sim 1/g^2$$



Adiabatic Quantum Optimization (AQO)

$$\mathbf{H} = (1 - s)\mathbf{H}_d + s\mathbf{H}_p$$

