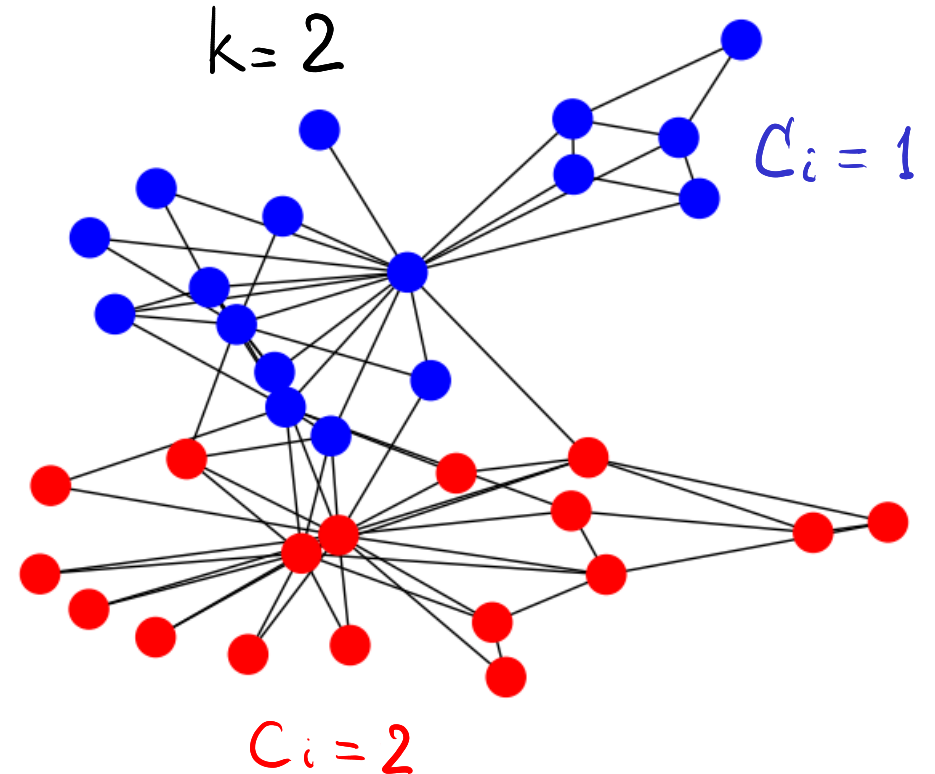
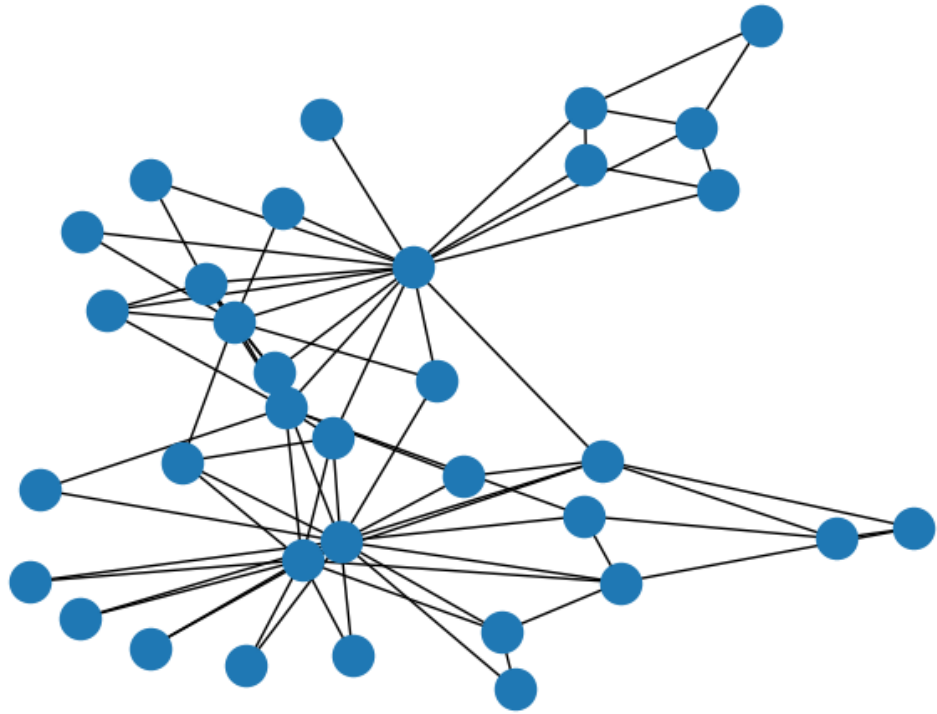


Community detection

on quantum
computers

About community detection

$$G : V, E \rightarrow C_1, C_2 \dots C_{|V|} \in \{1, \dots, k\}$$



Metrics

Modularity

$$M = \frac{1}{2m} \sum_{ij} \left(A_{ij} - \frac{g_i g_j}{2m} \right) \delta(c_i, c_j)$$

$$m = \frac{\sum_i g_i}{2}$$

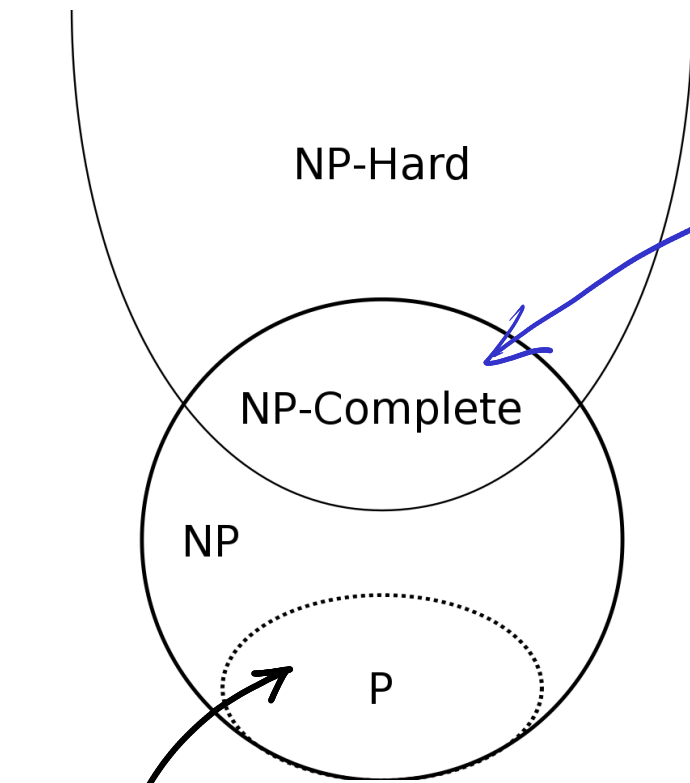
adjacency
matrix

Kronecker
delta

$$g_i = \sum_j A_{ij}$$

degree

The problem?



could be solved

We are here :(

For real-world graphs

- Annealing $\leftarrow t \rightarrow \infty$
- Greedy algorithms $\leftarrow M < M_{opt}$

[Submitted on 25 Aug 2006 (v1), last revised 30 Aug 2006 (this version, v2)]

Maximizing Modularity is hard

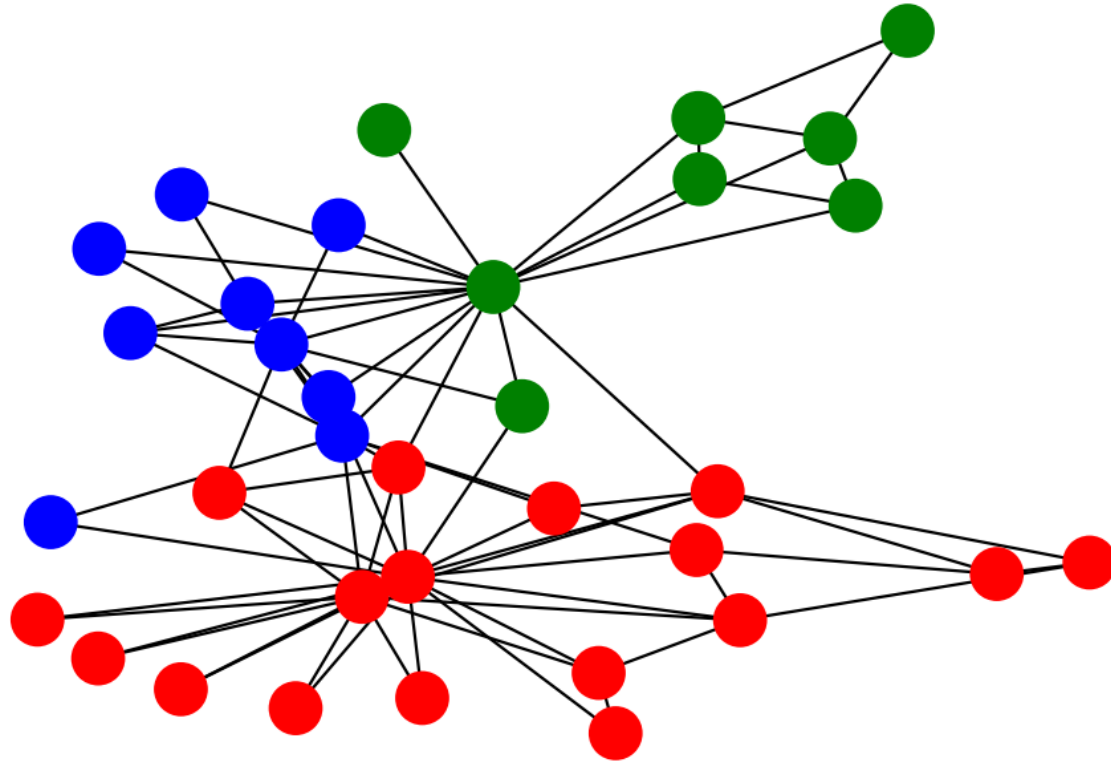
U. Brandes, D. Dellling, M. Gaertler, R. Goerke, M. Hoefer, Z. Nikoloski, D. Wagner

Several algorithms have been proposed to compute partitions of networks into communities that score high on a graph clustering index called modularity. While publications on these algorithms typically contain experimental evaluations to emphasize the plausibility of results, none of these algorithms has been shown to actually compute optimal partitions. We here settle the unknown complexity status of modularity maximization by showing that the corresponding decision version is NP-complete in the strong sense. As a consequence, any efficient, i.e. polynomial-time, algorithm is only heuristic and yields suboptimal partitions on many instances.

Comments: 10 pages, 1 figure

Subjects: **Data Analysis, Statistics and Probability (physics.data-an)**; Statistical Mechanics (cond-mat.stat-mech); Physics and Society (physics.soc-ph)

Greedy solution



$$M = 0.380$$

Clauset, A., Newman, M. E., & Moore, C. "Finding community structure in very large networks." Physical Review E 70(6), 2004.

Quantum stuff

Schr. eq.

$$i\hbar \frac{\partial}{\partial t} |\Psi\rangle = \hat{H} |\Psi\rangle$$

wave
function

Hamiltonian
(energy
operator)

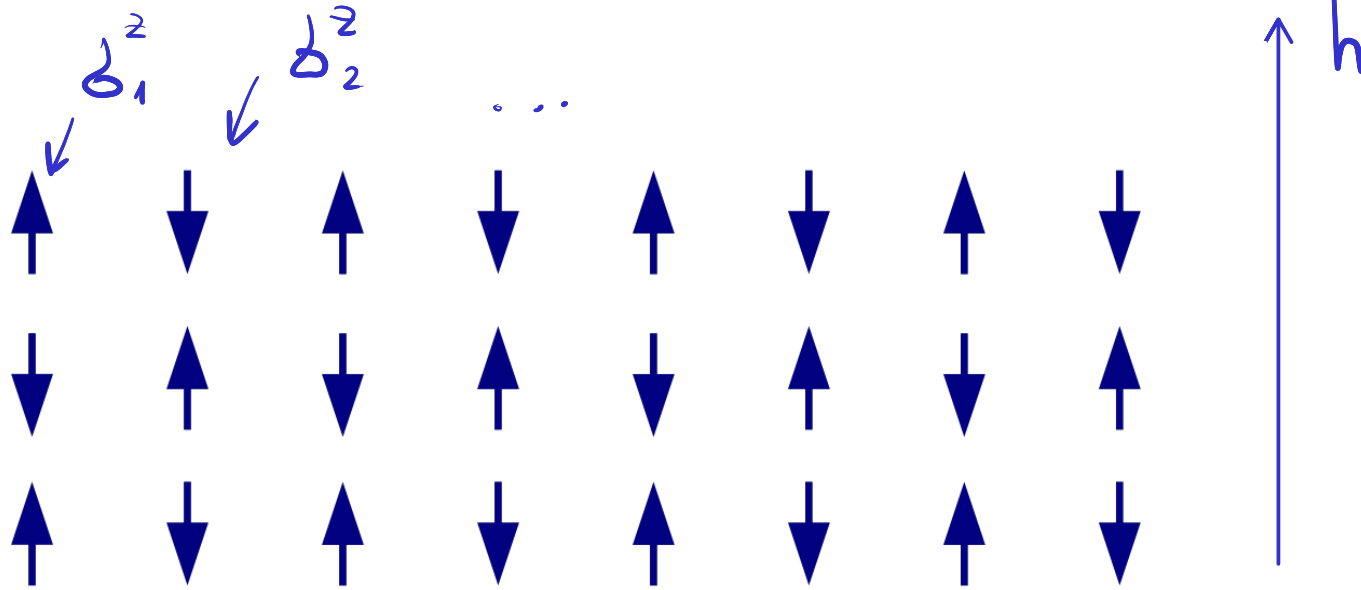
$$\hat{H} |\Psi_i\rangle = E_i |\Psi_i\rangle$$

Possible states
are eigenstates

„Ground state” —
— state with min. energy

$$E_i \geq E_{g.s.} = \langle \Psi_{g.s.} | \hat{H} | \Psi_{g.s.} \rangle$$

Ising model



Ernst Ising

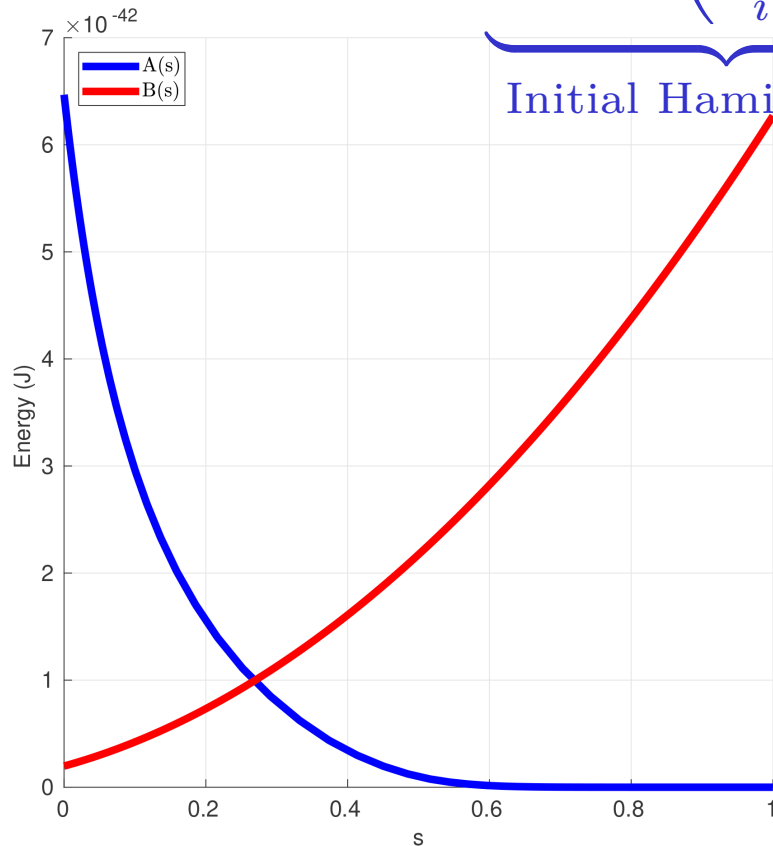
$$\hat{H} = -h \sum_i \sigma_i^z - J \sum_{i,j} \sigma_i^z \sigma_j^z$$

NP-hard

Has an effective
approximate q. alg.!

Quantum annealing |

$$\hat{H}_{Ising} = \underbrace{-A(t) \left(\sum_i \hat{\sigma}_x^{(i)} \right)}_{\text{Initial Hamiltonian}} + \underbrace{B(t) \left(\sum_i h_i \hat{\sigma}_z^{(i)} + \sum_{i,j} J_{i,j} \hat{\sigma}_z^{(i)} \hat{\sigma}_z^{(j)} \right)}_{\text{Final Hamiltonian}},$$



Adiabatic th.

$$\tau \rightarrow \infty$$

$$\hat{H}_0 \rightarrow \hat{H}_t$$

$$|\psi(0)\rangle \neq |\psi(t)\rangle$$

$\Psi(t)$ — eigenstate of \hat{H}_t !

Modularity as Ising

$$x_{i,k} = \begin{cases} 1, & \text{if } x \text{ in } c_k \\ 0, & \text{otherwise} \end{cases}$$

Cost part

$$\hat{H}_p = P \sum_i^N \left(\sum_c^k x_{i,c} - 1 \right)^2$$

Constraints

$$B_{i,j} = A_{i,j} - \frac{g_i g_j}{2m}$$

$$\hat{H}_M = -\frac{1}{m} X^T \begin{vmatrix} B & \dots & 0 \\ \dots & \dots & \dots \\ \dots & \dots & B \end{vmatrix} X$$

QUBO matrix

D-wave solution

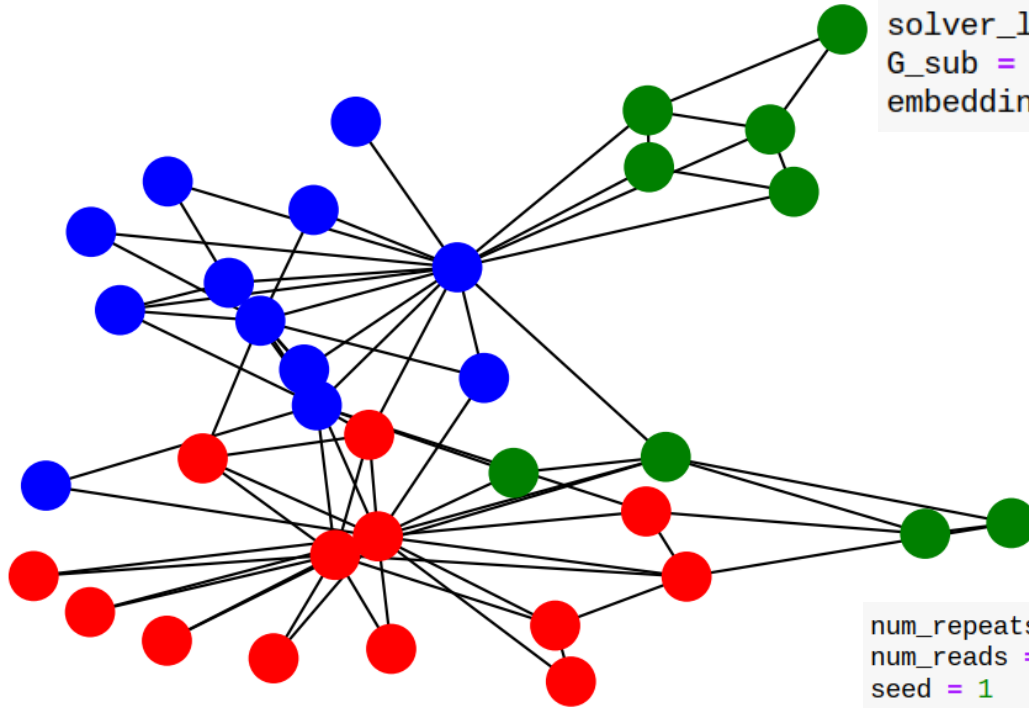
github.com/aws-samples/amazon-braket-community-detection

```
from dwave_qbsolv import QBSolv
from dwave.system import DWaveSampler, FixedEmbeddingComposite
```

```
import minorminer

solver_limit = 50
G_sub = nx.complete_graph(solver_limit)
embedding = minorminer.find_embedding(G_sub.edges, base_sampler.edgelist)
```

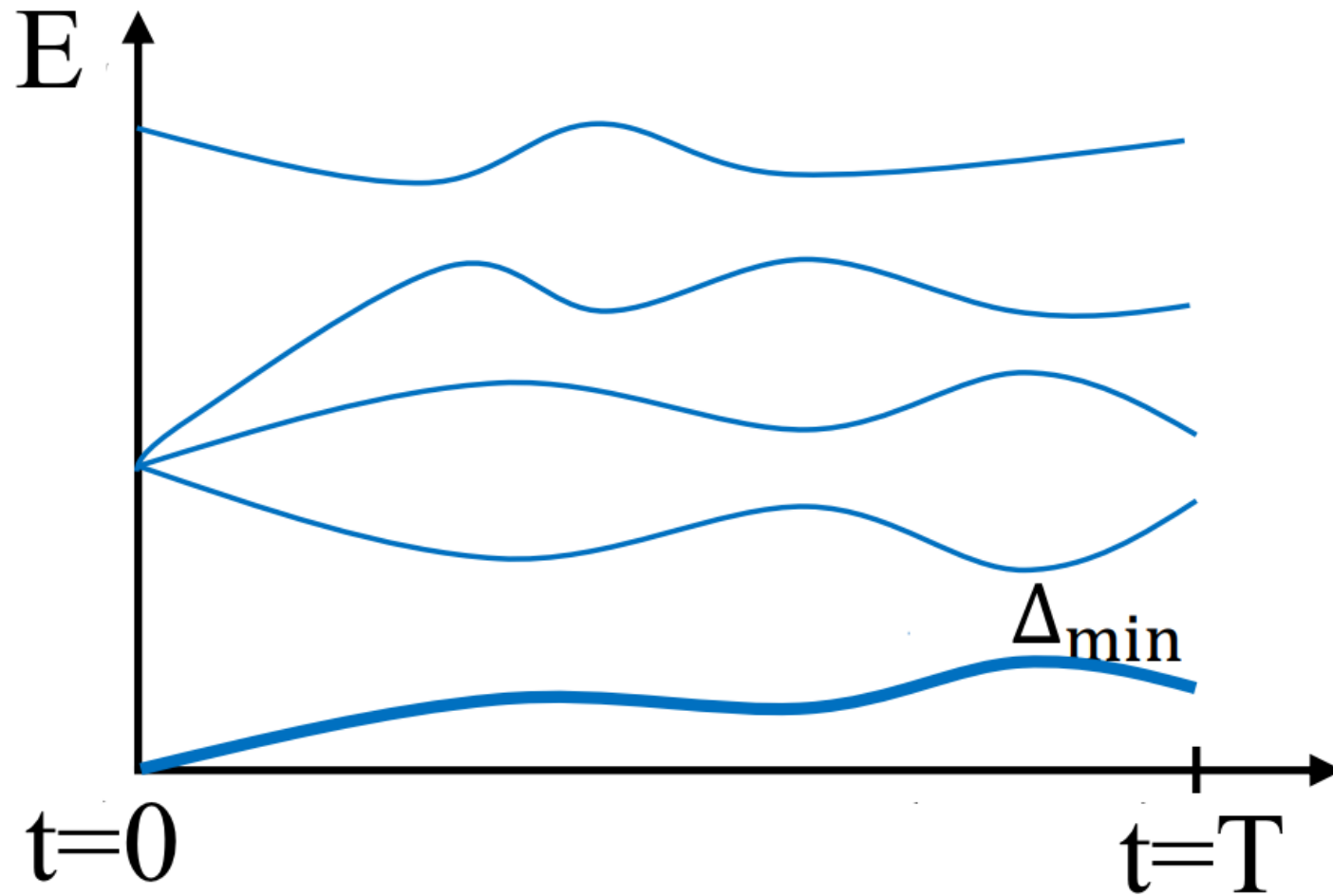
$$M = 0.396$$



```
num_repeats = 1
num_reads = 1000
seed = 1
response_hybrid = QBSolv().sample_qubo(qubo, solver=solver,
                                       num_repeats=num_repeats, solver_limit=solver_limit,
                                       num_reads=num_reads, seed=seed)
```

```
base_sampler = DWaveSampler(token=token)
solver = FixedEmbeddingComposite(base_sampler, embedding)
```

What's the catch?



$\Delta_{\min} > 0?$

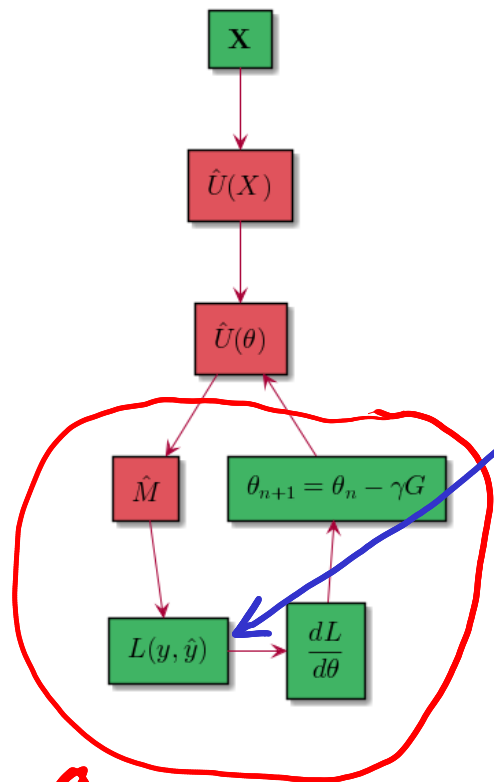
Another way?

VQE

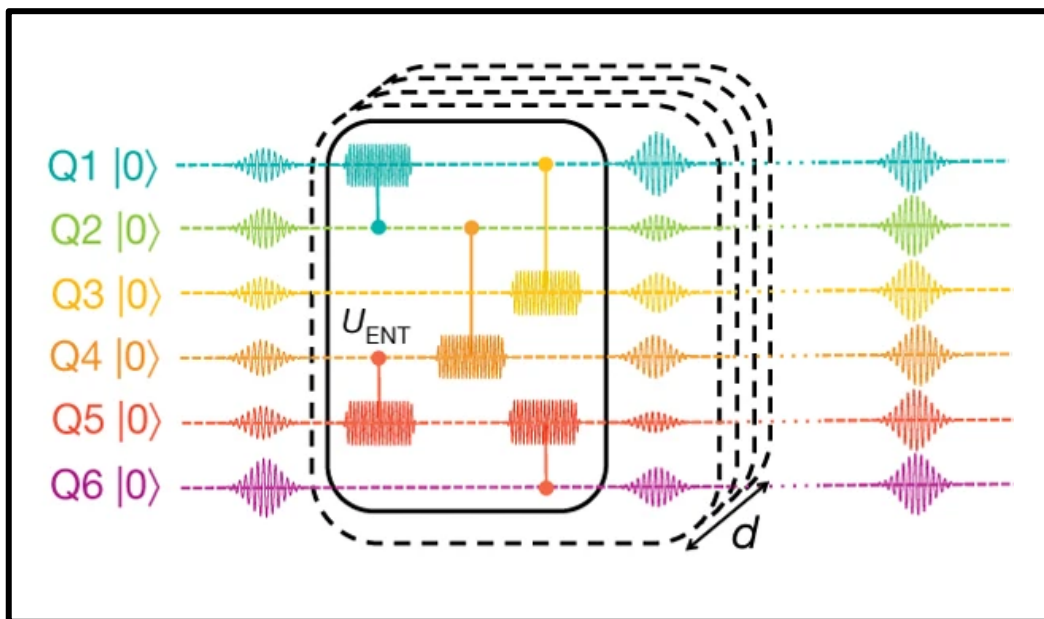
$$L(\theta) = \langle 0 | \hat{U}^*(\theta) H_{\text{Ising}} \hat{U}(\theta) | 0 \rangle$$

$$|\Psi\rangle \simeq \hat{U}(\theta) |0\rangle$$

Approximation



“Gradient descent”



Another way?

„Trotterization“

QAQA

$$|\Psi(t)\rangle = e^{-i\hat{H}_{Ising}(t)t}|\Psi(0)\rangle = e^{-i(A(t)\hat{H}_{initial}+B(t)\hat{H}_{cost})t}|\Psi(0)\rangle$$

Solution from Schr. eq.

$$\begin{aligned} A(t) &\rightarrow \gamma_1, \gamma_2, \dots, \gamma_N \\ B(t) &\rightarrow \beta_1, \beta_2, \dots, \beta_N \end{aligned}$$



$$|\Psi(t)\rangle = e^{-i\gamma_1\hat{H}_{initial}}e^{-i\beta_1\hat{H}_{cost}}\dots e^{-i\gamma_N\hat{H}_{initial}}e^{-i\beta_N\hat{H}_{cost}}|\Psi(0)\rangle$$

$$\arg \min_{\gamma_1, \dots, \gamma_N, \beta_1, \dots, \beta_N} \langle \Psi_{final} | \hat{H}_{cost} | \Psi_{final} \rangle$$

← Could
be solved
by SGD

Want to know
more?

