

## Potentials of Quantum Computing and First Applications in Computed Tomography

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**Abstract:** Quantum Computing (QC) technology has obtained a lot of attention recently.  
First QC-devices

are available as test beds for real-world applications. In this field, the Fraunhofer Gesellschaft  
has started a close collaboration with IBM in 2020. Since then, Fraunhofer Development  
Center

X-ray Technology (EZRT) has access to a real-world NISQ-device (Noisy Intermediate Scale  
Quantum), although the number of available qubits is still low (typically less than 100).

Basically, there are two types of QC-devices, the above mentioned NISQ-systems and the  
class

of Quantum Annealers which are restricted mostly to optimization problems, but today  
already

provide several thousands of qubits with each device. On the other hand, the NISQ-systems  
provide a certain set of “commands”, called quantum gates, which can be combined in any  
arbitrary way to form an elaborated quantum circuit. A quantum circuit may be seen as an  
analog to a program, although there are decisive differences. The Fraunhofer EZRT is  
involved in three projects concerning quantum computing, dealing with

general research questions in this field, but also specifically with industrial research activities.  
In them, we research and apply existing quantum computing solutions to numerical problems  
as they often occur in the field of X-ray Computed Tomography (CT), as well as  
experimentally

determine practical limits to the existing theoretical works.

In our talk we review the current state of QC-technology and give a short overview of the  
theoretical background. The concept of qubits, superposition, entanglement and measurement  
of quantum states will be explained. Quantum gates and circuits built out of these gates will  
be

introduced. In the second part, we will report on the current state of our research. The three  
running

projects include an implementation of cross-sectional image reconstruction for CT imaging  
based on a QUBO-method (Quadratic Unconstrained Binary Optimization), used to find a  
mathematical minimum of a metric.

Another topic, we are working on, is optimization of X-ray source trajectory in CT with respect  
to image quality achievable by a restricted number of angular positions. In addition, we  
evaluate common problems of image processing for instance noise reduction and search for  
ways to implement respective methods on the QC. Acknowledgement

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# Potentials of Quantum Computing and First Applications in Computed Tomography



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# Overview

## Computed Tomography and Quantum Computing

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### Introduction (1)

- **Principles of Quantum Computing (QC)**
- **Programming a QC-device**

### Introduction (2)

- Principles of Computed Tomography (CT)

### Putting things together

- QC-based optimization of trajectories
- QC-based Image reconstruction

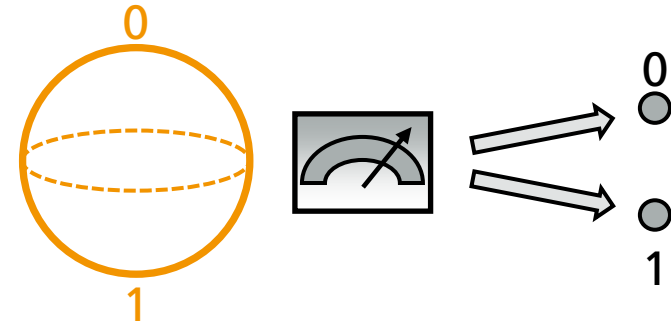
### Conclusions & Outlook

# QC theory

What is superposition and what does a measurement?

Measurement in quantum physics:

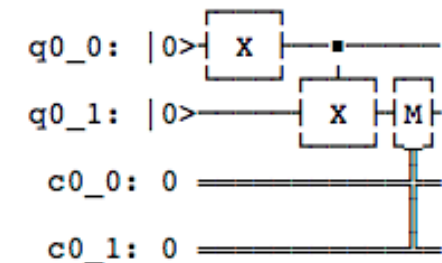
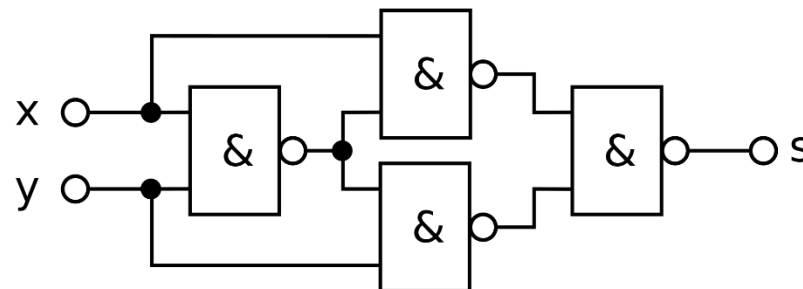
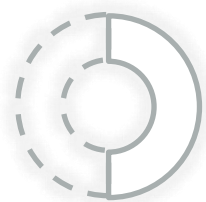
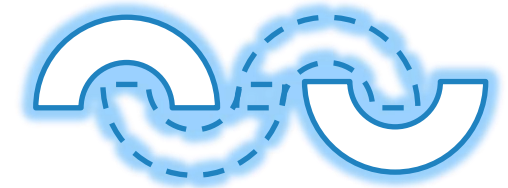
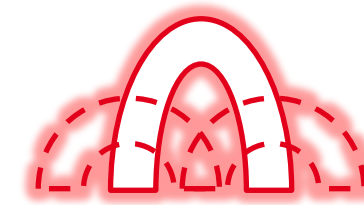
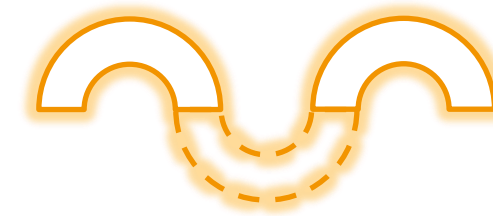
- When a qubit is read out (measured), the result is 0 or 1
- Coefficients  $c_0$  und  $c_1$  in  $\begin{pmatrix} c_0 \\ c_1 \end{pmatrix}$  represent the probability:
  - Probability for measuring 0:  $|c_0|^2$
  - Probability for measuring 1:  $|c_1|^2$
- Normalization:  $|c_0|^2 + |c_1|^2 = 1$
- After measurement the qubit is in the measured state (called collapse of the wave function)



# QC theory

## Similarity of classical and quantum computers

- Specific input state  
QC: with **superposition**
- Concatenation of **gates**  
QC: Qubits get manipulated and connected to each other  
→ **interference** and **entanglement**
- Read (qu)bit by measurement  
QC: several measurements are necessary because result depends on probability



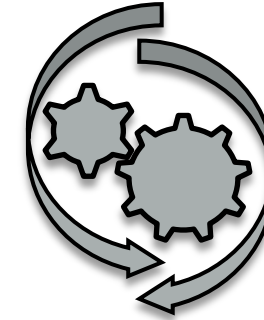
# QC theory

## Differences between classical and quantum computers

### QC gates must be invertible

As many outputs as inputs

Mathematical: matrix which describes gate must be unitary



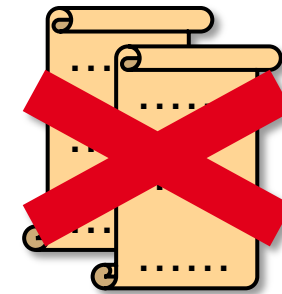
### Qubit states cannot be copied (no-cloning-theorem)

„result state“ at the end of a QC circuit cannot be copied

Problem: state is „destroyed“ by measurement (wave collapse)

Several executions are needed to get the distribution of the result

→ 100 up to 1000 “shots”, histogram of probability for all possible results





# Tensor product

A **quantum register** is a combination of several qubits

→ mathematically denoted as tensor product:  $\otimes$

$$\begin{pmatrix} \beta_0 \\ \beta_1 \end{pmatrix} \otimes \begin{pmatrix} \gamma_0 \\ \gamma_1 \end{pmatrix} = \begin{pmatrix} \beta_0 \gamma_0 \\ \beta_0 \gamma_1 \\ \beta_1 \gamma_0 \\ \beta_1 \gamma_1 \end{pmatrix}$$

Analogous **quantum gates**, example Hadamard:

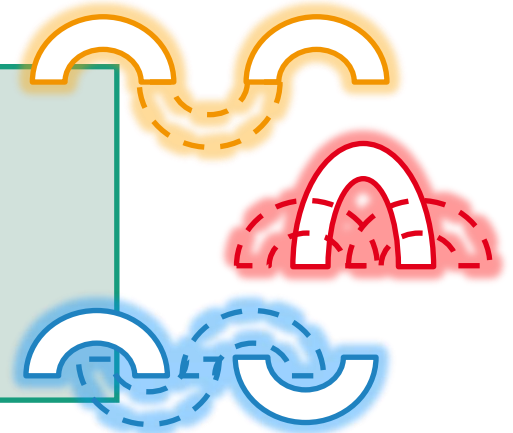
$$H \otimes H = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \otimes \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

# QC – basic facts

## What does make quantum computers so powerful?

Why is QC (sometimes) more powerful than classical computing?

- Use of quantum mechanical properties: **superposition**, **interference** and **entanglement**
- QC can be faster than classical computers, but not in general
- Special algorithms are needed, which are difficult to develop



Example: Memory usage for the simulation of a quantum computer

- Representation of one qubit as two complex numbers represented as four real 8 byte numbers
- Memory usage for one qubit:  $4 \cdot 8 \text{ byte} = 2^5 \text{ byte}$
- Each additional qubit doubles the memory usage!
- Simulation of a QC with 45 qubits:  
Memory usage:  $2^{44} \cdot 2^5 \text{ byte} = 5.62949953421 \cdot 10^{14} \text{ byte} = 0.5 \text{ petabyte}$

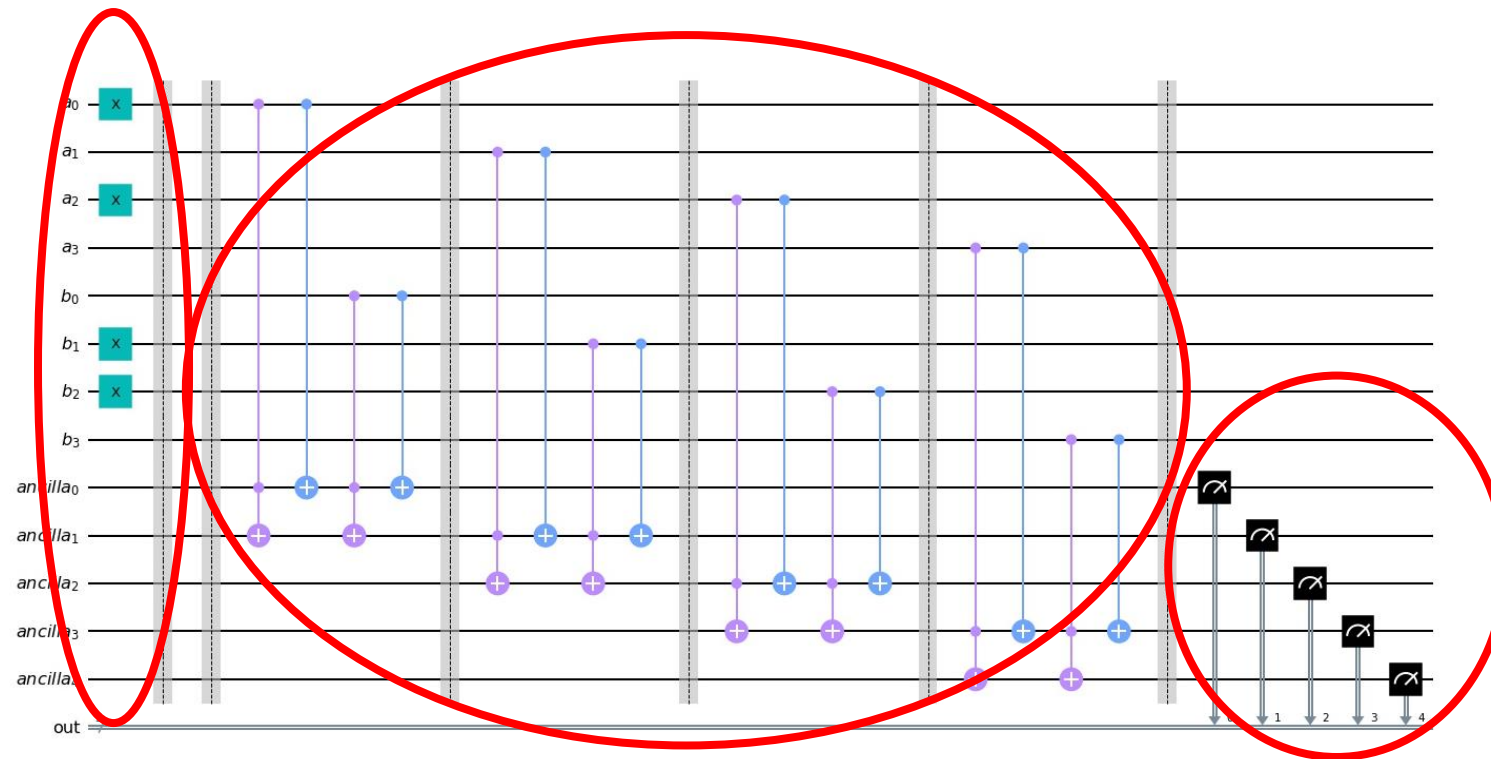


# How to write a „program“ on a QC-device?

prepare qubits

calculation

measurement



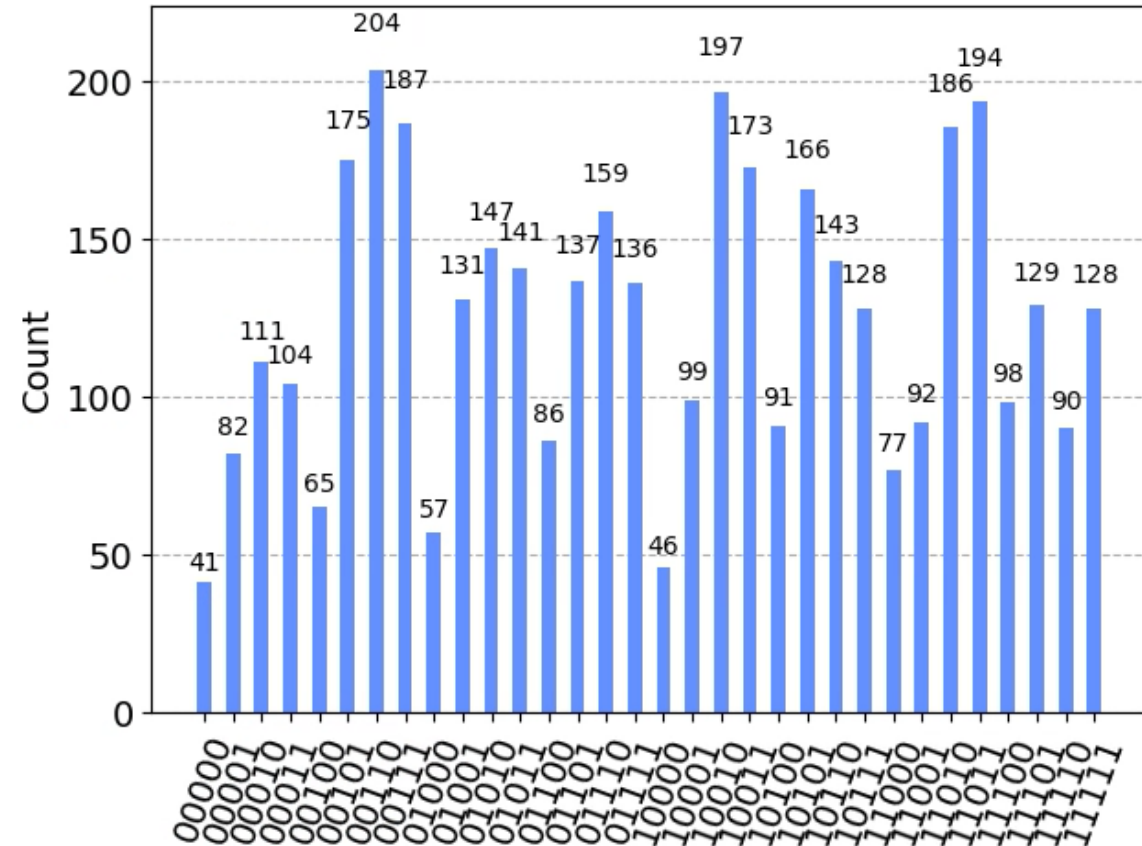
# How to retrieve results from a QC-device?

## → INTERPRETING HISTOGRAMS

Repeat execution of quantum circuit many times

Typical 100 ... 1000 “shots”

Resulting in a Histogram showing the frequency of occurrence of all possible states



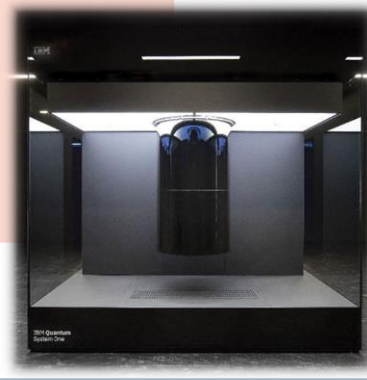
# Quantum Computing – some basic facts

## Types of QC hardware

### Universal Quantum Computer

- arbitrary quantum circuits by stringing together gates that manipulate qubits (quantum bits)
- various realizations:
  - ion traps:  
Manipulation of single ions = qubit (encoding via energy states)
  - optical systems:  
photon = qubit (coding e.g. via polarization)
  - superconductors with a Josephson contact
  - and other technologies...

> 127 qubits



### Quantum Annealer

- lattice-like networking of qubits
- qubits cannot be manipulated individually
- can only solve special problems  
→ not a universal computer
- solution of minimization problems  
→ problem must be formulated in this way
- best known quantum annealer:  
D-Wave (qubits based on superconductivity)

> 5000 qubits



# Overview

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### Putting things together

- QC-based optimization of trajectories
- QC-based Image reconstruction

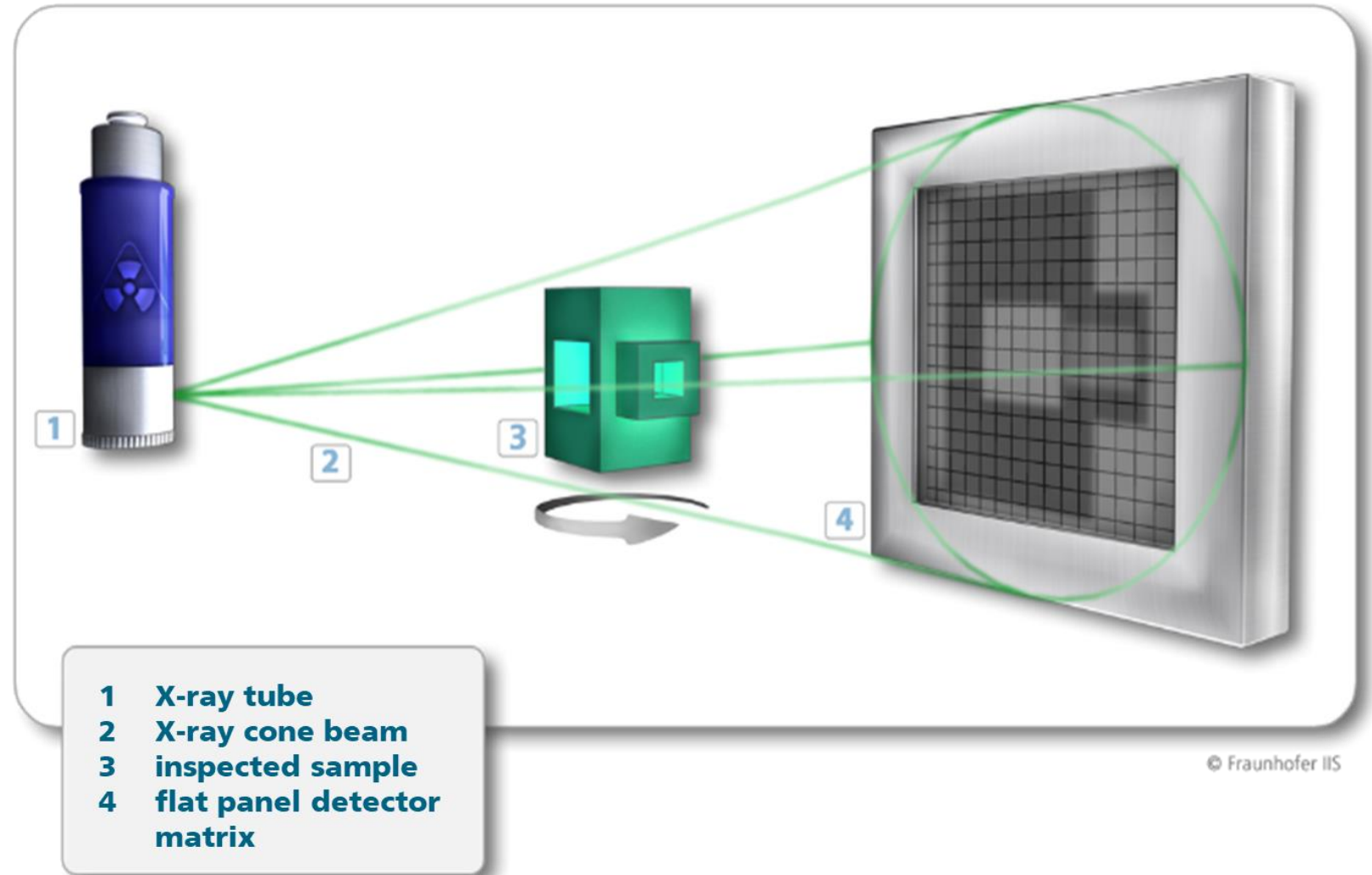
### Conclusions & Outlook

# Optimization of CT Data Acquisition by means of Quantum Computing

## Principles of Computed Tomography Imaging

### Conventional data acquisition process

- move X-ray source on a circular path around object
- equivalent: rotate object between X-ray source and flat panel sensor
- complete 360° data set
- simple parametrization
- drawback: time consuming, redundant information



# Optimization of CT Data Acquisition by means of Quantum Computing

## Optimizing Trajectories

### Why using non-circular trajectories?

- move x-ray source and detector array independently by robots
- improve access to complex and / or very large parts
- save time & money by accessing only the “most important” projections, e.g. reducing from 1000 to 5000 views to 100 to 200
- avoid artifacts by eliminating problematic directions



# Overview

## Computed Tomography and Quantum Computing

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### Introduction (1)

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### Putting things together

- **QC-based optimization of trajectories**
- **QC-based Image reconstruction**

### Conclusions & Outlook



# Optimization of CT Data Acquisition by means of Quantum Computing

## Optimizing Trajectories for Computed Tomography

### Formulation of the optimization problem

- **vector  $\vec{p}$  is defined for each projection:**
  - direction encodes angular position of the X-ray source (from where the projection is created)
  - length depicts the edge quality of the object from that direction
- **the basic idea is, do NOT search for the  $n$  “best” out of  $N$  projections, but for the “best” set of  $n$  projections**
- **just looking at single projection by measuring their information content, will lead to very similar data from close-by positions**
- **several metrics were developed and tested**

$$\vec{p}_i = \sum_{j=1}^J c_{j,i}^{max} \cdot \frac{(\vec{d}_{j,i} - \vec{s}_i)}{|\vec{d}_{j,i} - \vec{s}_i|}$$

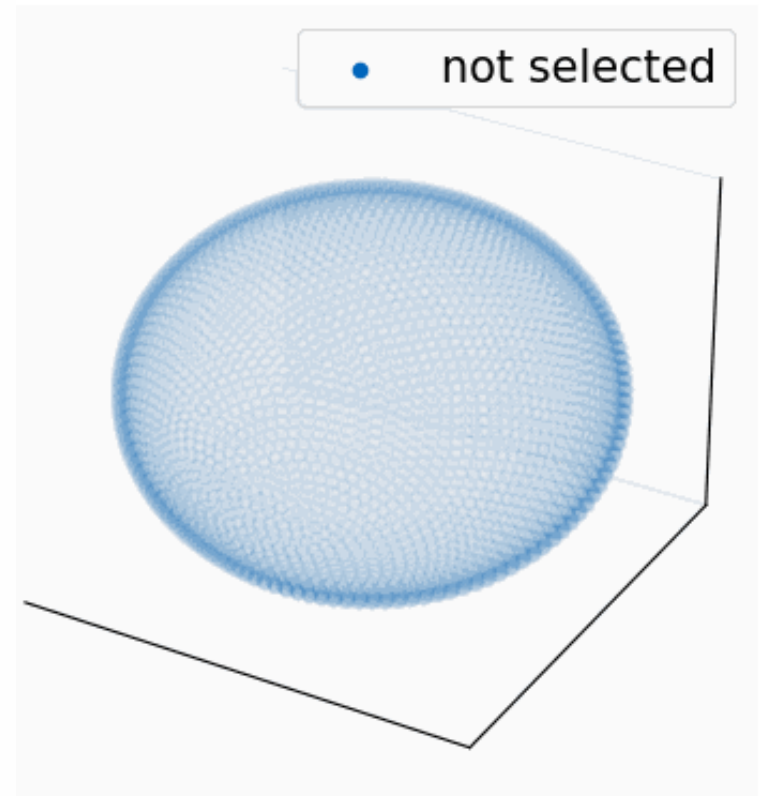
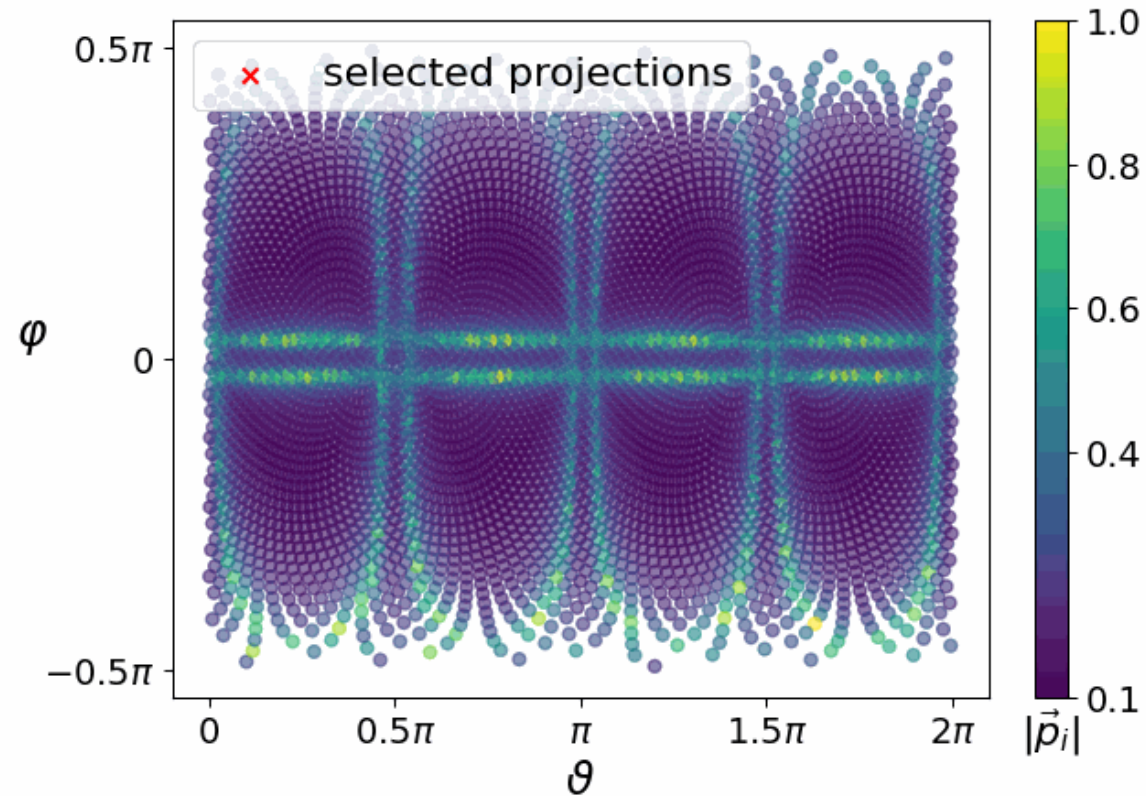
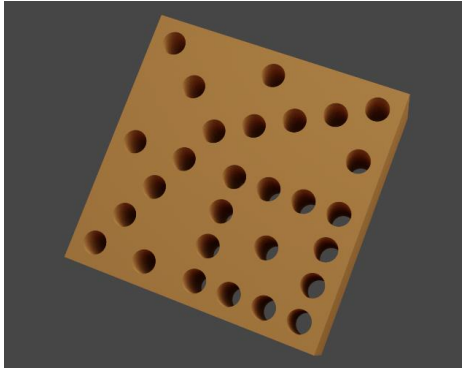
$c_{j,i}^{max}$ : **max. wavelet coefficient**

$\vec{d}_{j,i}$ : **corresponding pixel position**

$\vec{s}_i$ : **source position**

# Partial metric 1: Reducing redundancy due to neighboured projections

## Optimizing Trajectories for Computed Tomography



→ maximization of edge information from many directions

# Quantum Computing – well known applications

## Optimization by QC

### Use Quantum Annealer to solve combinatory binary optimization problems

- these mathematical problems include:
  - **Q**uadratic **B**inary **O**ptimization
  - Ising-models
  - graph theory
- the approaches can be transformed into each other:
  - graph problems can be reformulated as QUBOs
  - with  $s = 2x - 1$  the QUBO becomes an Ising-model

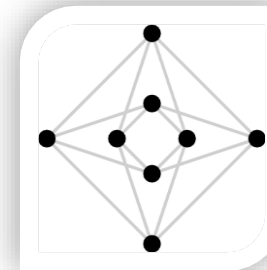
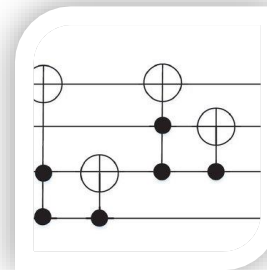
**QUBO**

$$f_Q(x) = \sum_{i=1}^n \sum_{j=1}^i q_{ij} x_i x_j \quad x_i = \{0, 1\}$$

**Ising-model**

$$\hat{\mathcal{H}} = -\frac{1}{2} \sum_{i,j} J_{ij} s_i^z s_j^z - H_z \sum_{i=1}^N s_i^z$$

$s_i^z = \pm 1$



# Optimization of CT Data Acquisition by means of Quantum Computing

## Putting everything together

- set-up model for X-ray measurement
- test object: perforated plate
- calculate virtual projections
- define measure to quantify „amount of information“ in each projection
- combine 2 metrics

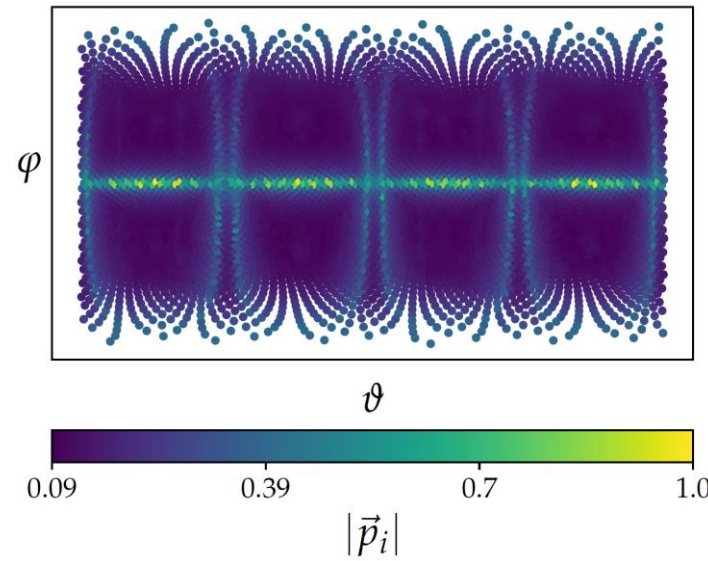
$$f_{M1}(x) = \sum_k^3 \left\| \sum_i^N p_{k,i} x_i \right\|^2$$

$$f_{M2}(x) = \sum_{i,j}^N V_{i,j} x_i x_j$$

$$f_Q(x) = f_{M1}(x) + f_{M2}(x),$$



simulated projection of the plate at  $\phi = 0^\circ$  and  $\vartheta = 0^\circ$  (w/o photon noise)



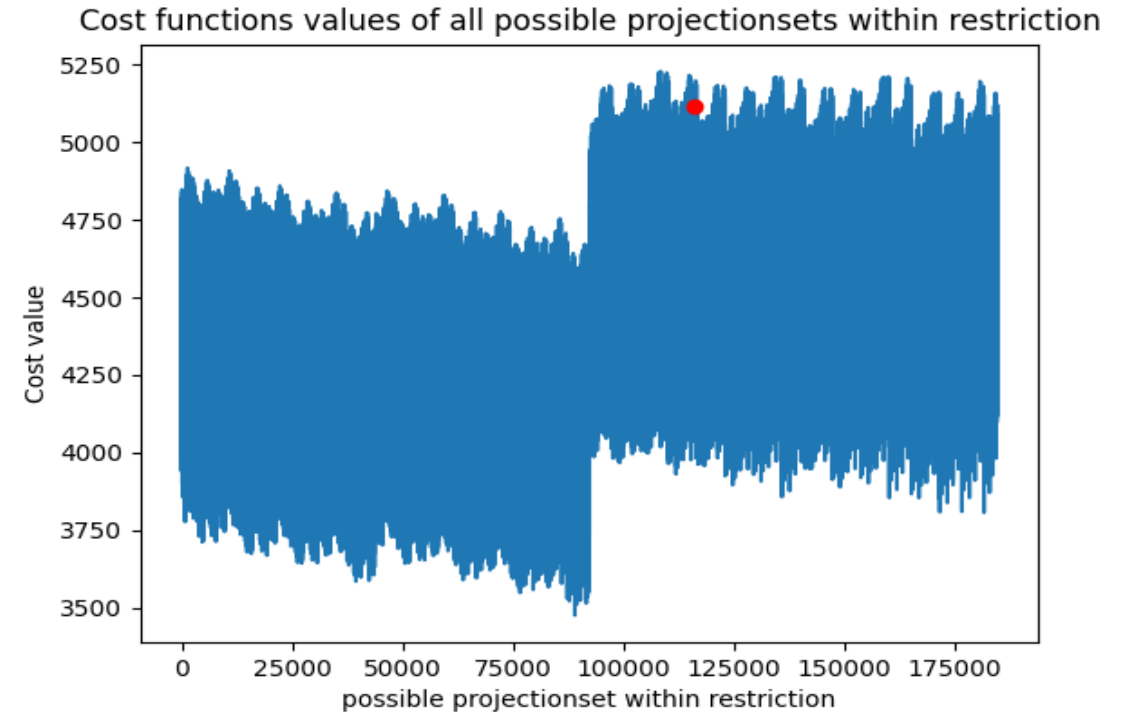
map of quality measure  $\mathbf{p}$  for 5,000 projection equally distributed on a  $4\pi$  solid angle

# Optimization of CT Data Acquisition by means of Quantum Computing

## Results

### Cost function values of all possible projection sets

- $x_i = 1$  means projection  $p_i$  is taken into account for reconstruction, else not ( $x_i = 0$ )
- Chose 10 out of 20 projection sets
- there are  $\binom{20}{10} = 184756$  possible combinations
- the ordinate shows a label given to all sets of projections taken into account  $\rightarrow$  variable  $x$
- the red dot marks the result from the QC



$$\{x_1, \dots, x_{20}\}$$

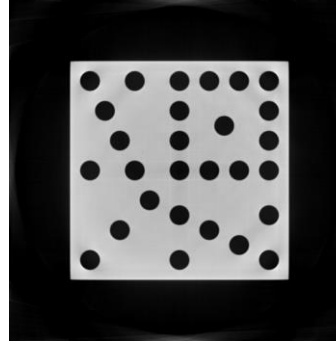
# Optimization of CT Data Acquisition by means of Quantum Computing

## Results

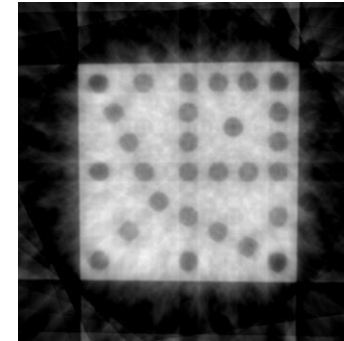
### Cross-sectional images of test body

- ground truth vs. considerably reduced number of projections
- reconstruction from 10 selected projections on a semi-circular trajectory

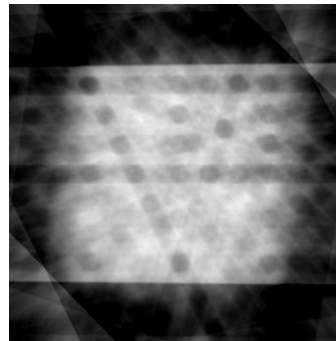
1,000 equidistant projections



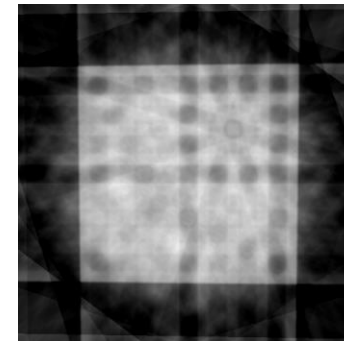
20 equidistant projections along a semi-circular trajectory



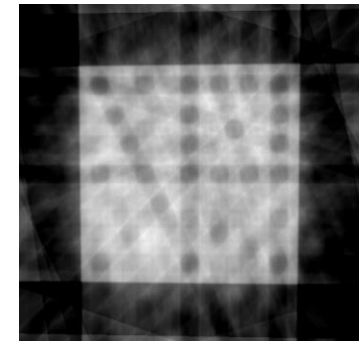
10 randomly sampled projections



10 equidistant projections



projection set from QC



# QUBO based Ansatz for CT image reconstruction

Working on a discrete representation of the data

## Formulation of the task

- Reconstruction problem = inverse problem

$$y = Mx$$

$y \rightarrow$  measured data (projections)

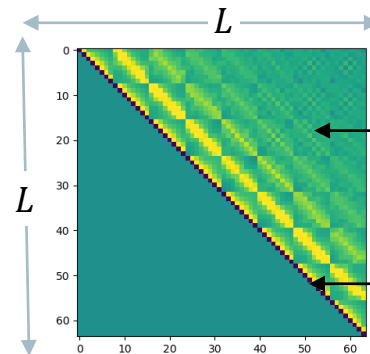
$M \rightarrow$  matrix describing the CT system

$x \rightarrow$  reconstructed volume with  $L$  pixels

- Least squares formulation:

$$H(x) = \|Mx - y\|_2^2 = \underbrace{x^T M^T M x}_{\text{quadratic in } x} - \underbrace{x^T M^T y - y^T M x}_{\text{linear in } x} + \underbrace{y^T y}_{\text{constant}}$$

- QUBO size:  $L^2$



$$Q_{i,j} = \sum_k M_{ki} M_{kj}$$

$$Q_{i,i} = -2 \sum_j y_j M_{ji}$$

negligible

Nau MA et al, Exploring the Limitations of Hybrid Adiabatic Quantum Computing for Emission Tomography Reconstruction. *Journal of Imaging*. 2023



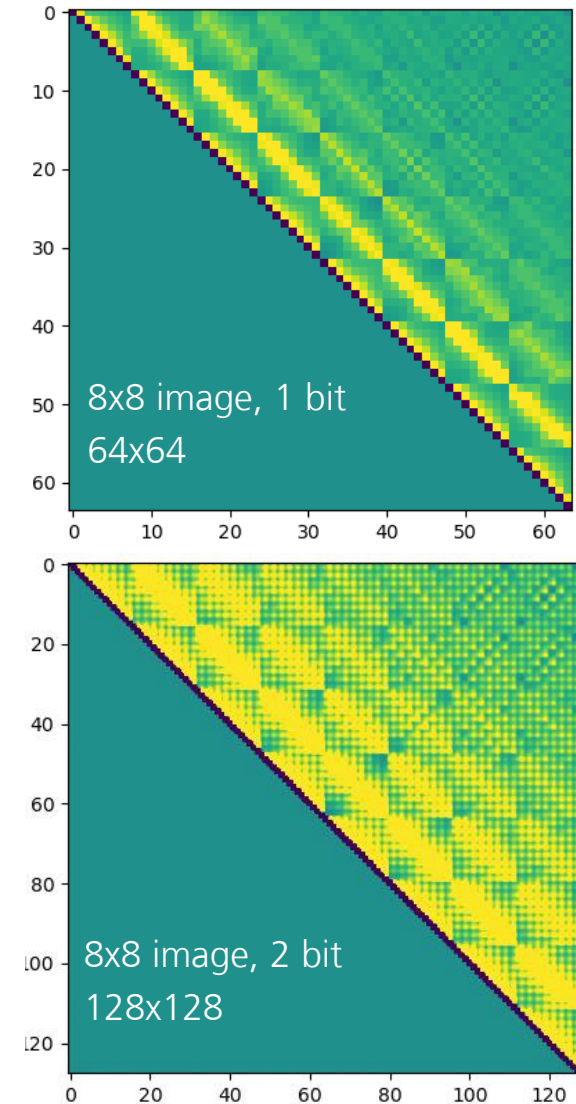
# QUBO based Ansatz for CT image reconstruction

## Multi bit representation

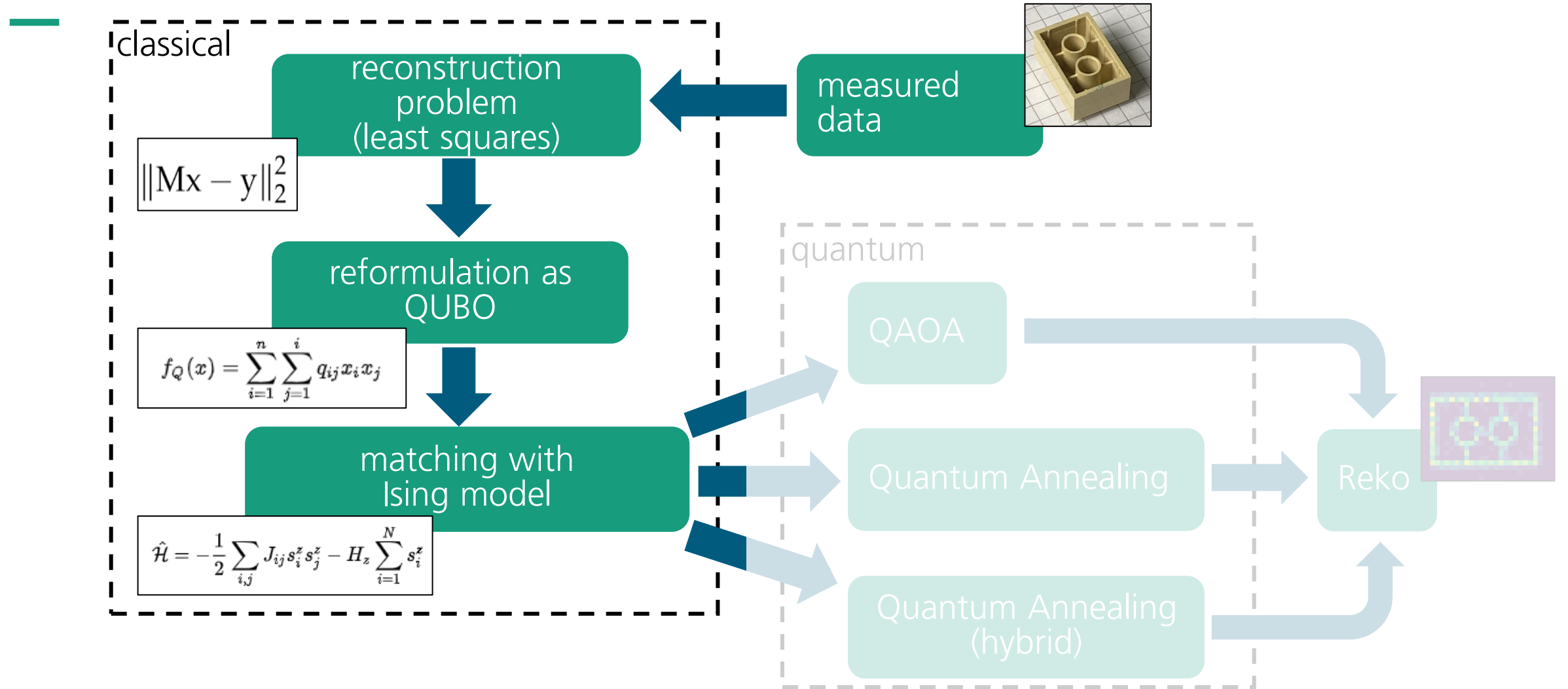
- Replace variable  $x_i$  for n-bit-representation by n binary variables

$x_{i,0}, x_{i,1}, \dots, x_{i,n-1}$  in binary representation

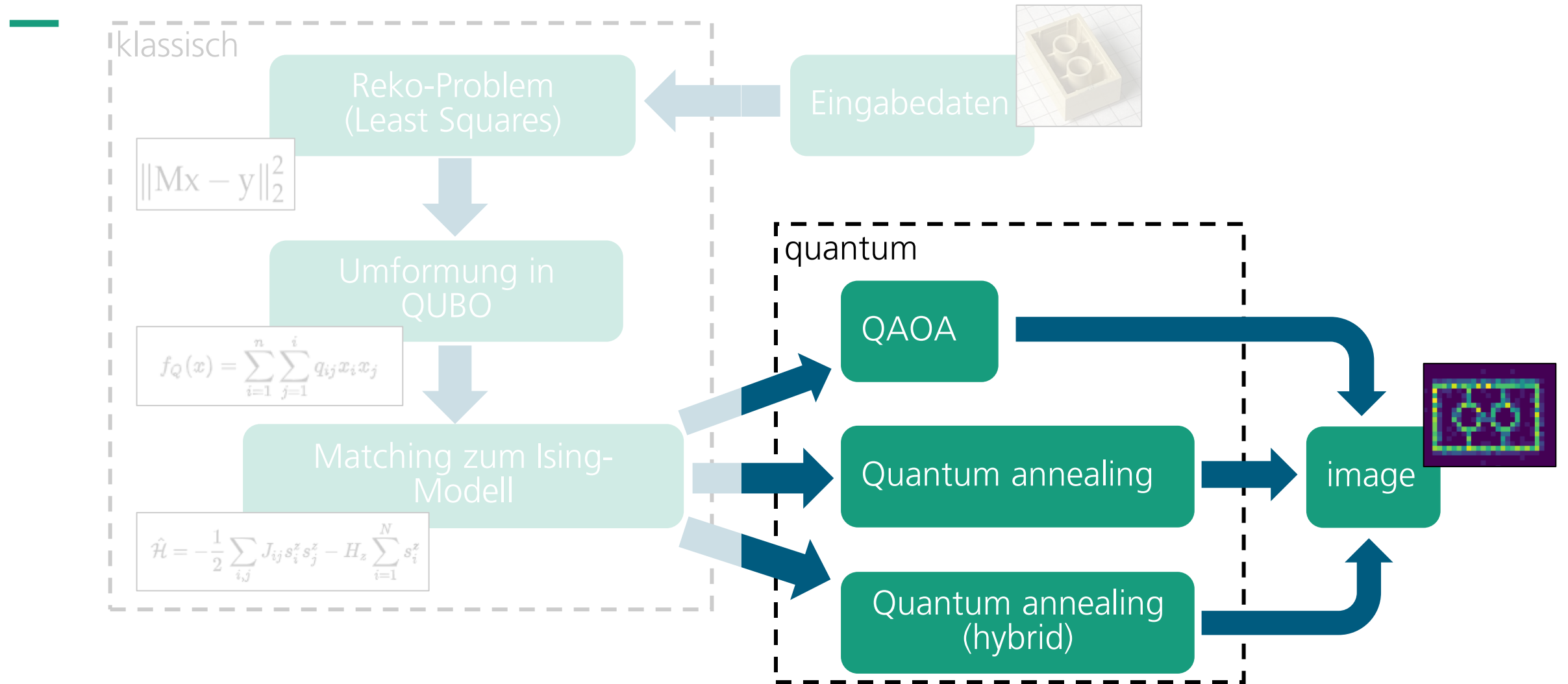
- Size of the QUBO matrix  $(L * n)^2$
- Examples for different dimensions:
  - Reconstruction 64 by 64 pixels @ 1 bit  
→ 16,7 Mio. combinations
  - Reconstruction 64 by 64 pixels @ 4 bit  
→ 267 Mio. combinations
  - Reconstruction 1024 by 1024 pixels @ 4 bit  
→  $17,6 * 10^{12}$  combinations



# QUBO based Ansatz of CT image reconstruction



# QUBO based Ansatz of CT image reconstruction

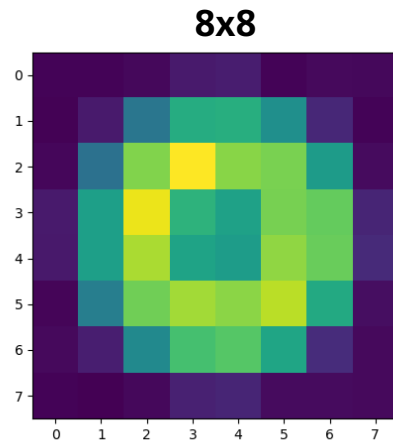


# Working on real (measured) data

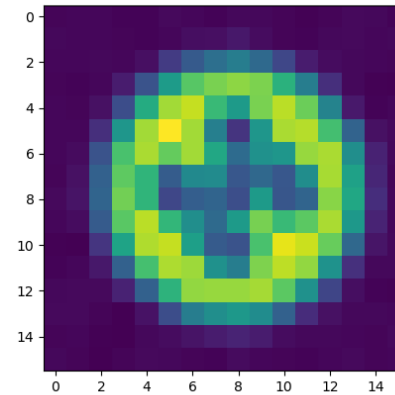
real data – conventional filtered back projection FBP

- $\mu$ -CT, 60 kV
- #projections 1200
- angular range: 360°

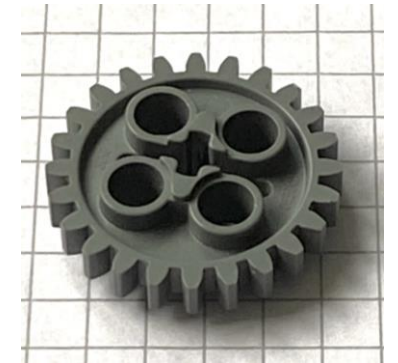
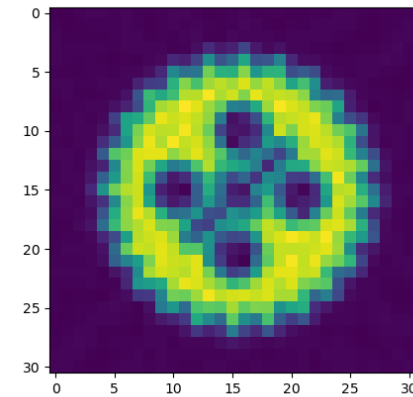
Lego toothed wheel



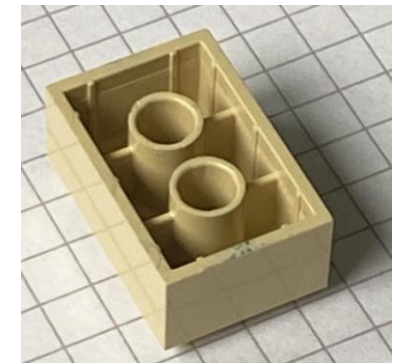
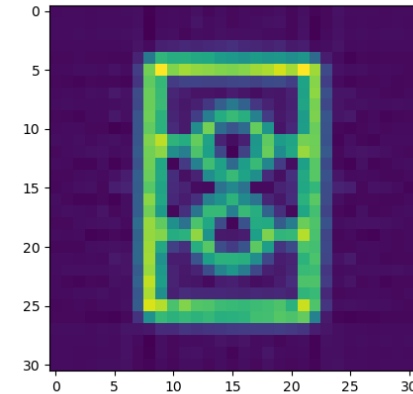
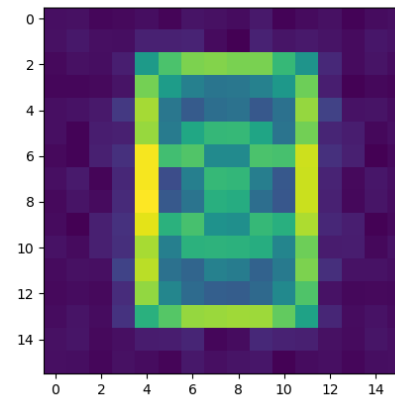
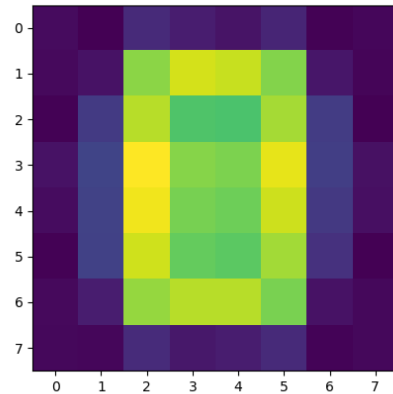
**16x16**



**32x32**



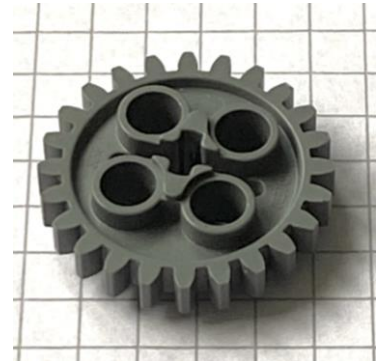
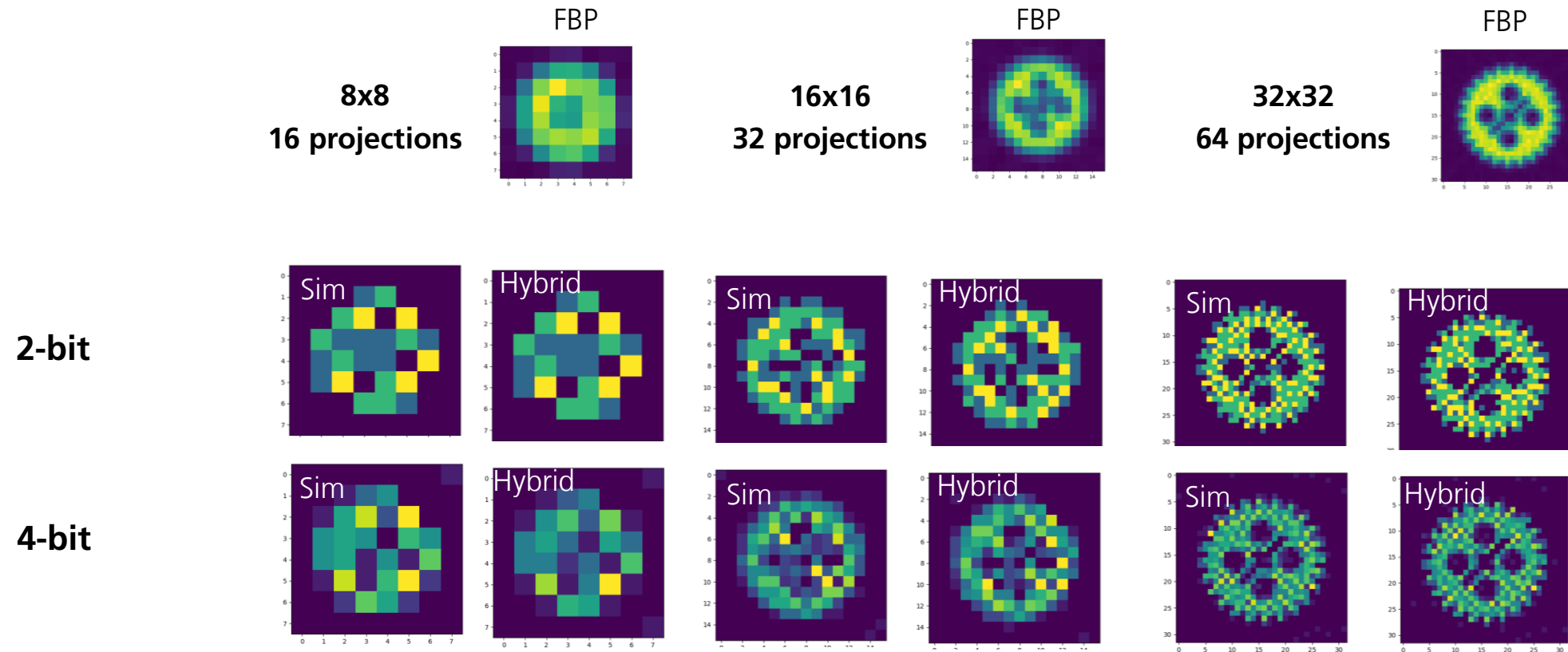
Lego brick



# Working on real (measured) data

## Simulated annealing and hybrid quantum annealing

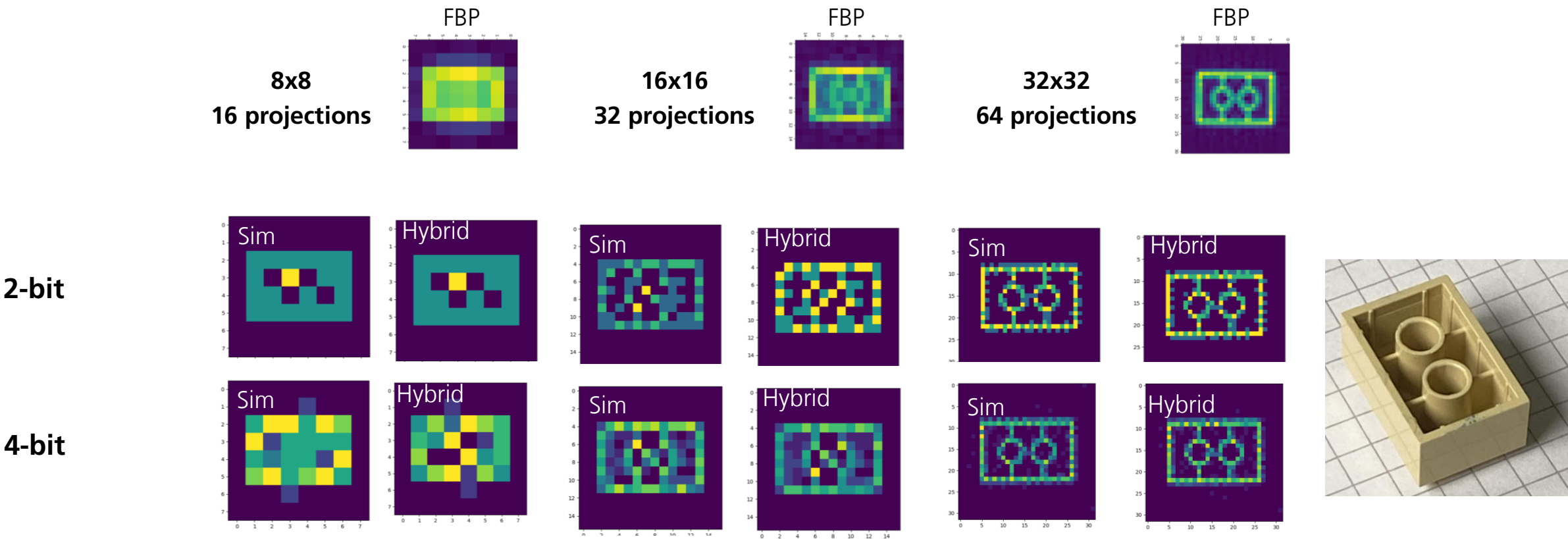
### Lego toothed wheel



# Working on real (measured) data

## Simulated annealing and hybrid quantum annealing

### Lego brick



# Optimization of CT Data Acquisition by means of Quantum Computing

## Summary & Discussion

### Optimization of few projections trajectory

- we developed & implemented an algorithm to optimize the selection of trajectories for Computed Tomography on a real-world Quantum Computer
- the stochastic complexity of the task increases quickly with higher numbers of projections.

Thus, to solve the task of optimization via a QUBO algorithm on a QC is expected to show advantages of quantum computing for a real-world problem

### CT image reconstruction

- image reconstruction for industrial CT was performed on a QC-device. Nevertheless, today image size and pixel dynamics are still very limited.

We expect to improve the QC-devices rapidly – reaching usual image dimensions in foreseeable time

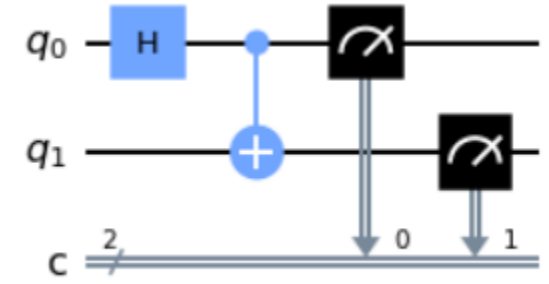
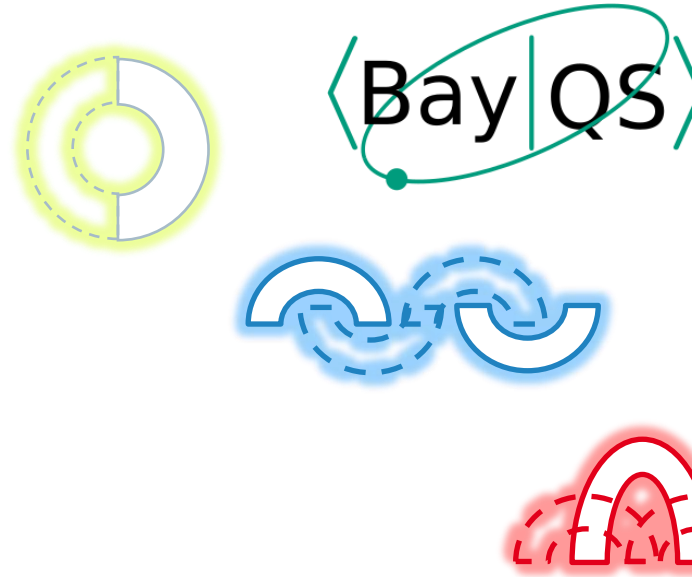
**Research on these topics will continue...!**



# Optimization of CT Data Acquisition by means of Quantum Computing

The End

Thank you for your  
attention!



11th Conference on Industrial Computed Tomography, Wels, Austria (iCT2022):

R. Schielein et al. „Quantum Computing and Computed Tomography: A Roadmap towards QuantumCT“

12th Conference on Industrial Computed Tomography, Fürth (iCT2023):

T. Lang, S. Semmler, et al. „N-Dimensional Image Encoding on Quantum Computers »

13th ECNDT 2023 - European Conference on Non-Destructive Testing (ECNDT2023), Lissabon July 3rd – 5th, 2023:

T. Fuchs et al. „Optimization of Computed Tomography Data Acquisition by means of Quantum Computing“

13th Conference on Industrial Computed Tomography (iCT2024), February, 6th- 9th 2024, Wels:

D. Prjamkov et al. „Comparison of Different Quantum Computing Devices for Optimization of Computed Tomography Data Acquisition“