

# Identification and Mitigation of Biases in Quantum Machine Learning

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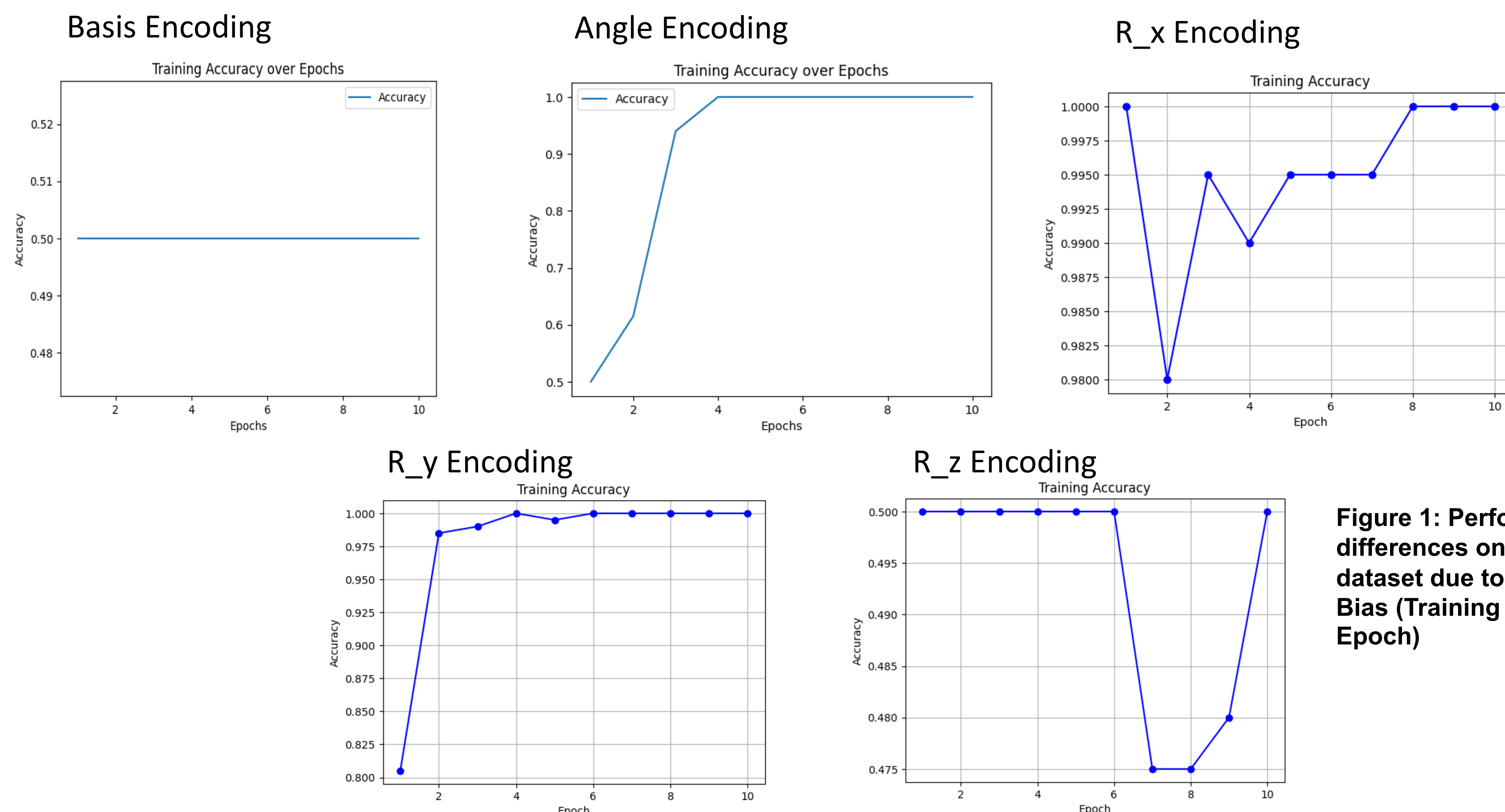
**Summary:** We identify biases in Quantum Machine Learning (QML) that arise from the distinct properties of quantum systems, which differ from biases in classical machine learning and review existing mitigation strategies.

## Featured Examples

### Encoding Bias

**PROBLEM:** Encoding Bias arises from the interaction between the transformation of classical data into quantum states, and the quantum algorithm.

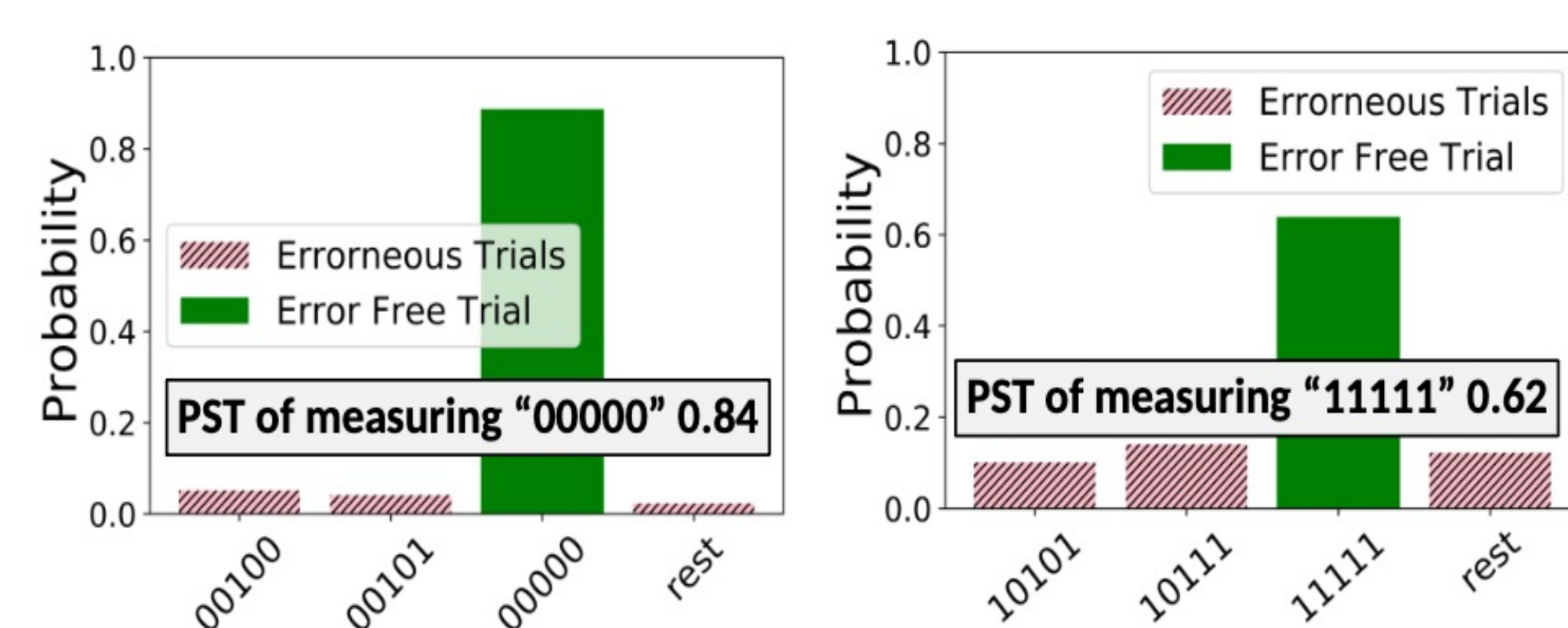
**OUR EXPERIMENT:** To examine this bias, we conducted an experiment using various encoding techniques applied to a fixed QNN architecture on the MNIST dataset for classification tasks. Our results (see below) show the significant impact of encoding choice on model performance



**Figure 1: Performance differences on MNIST dataset due to Encoding Bias (Training accuracy vs Epoch)**

### State-Dependent Bias

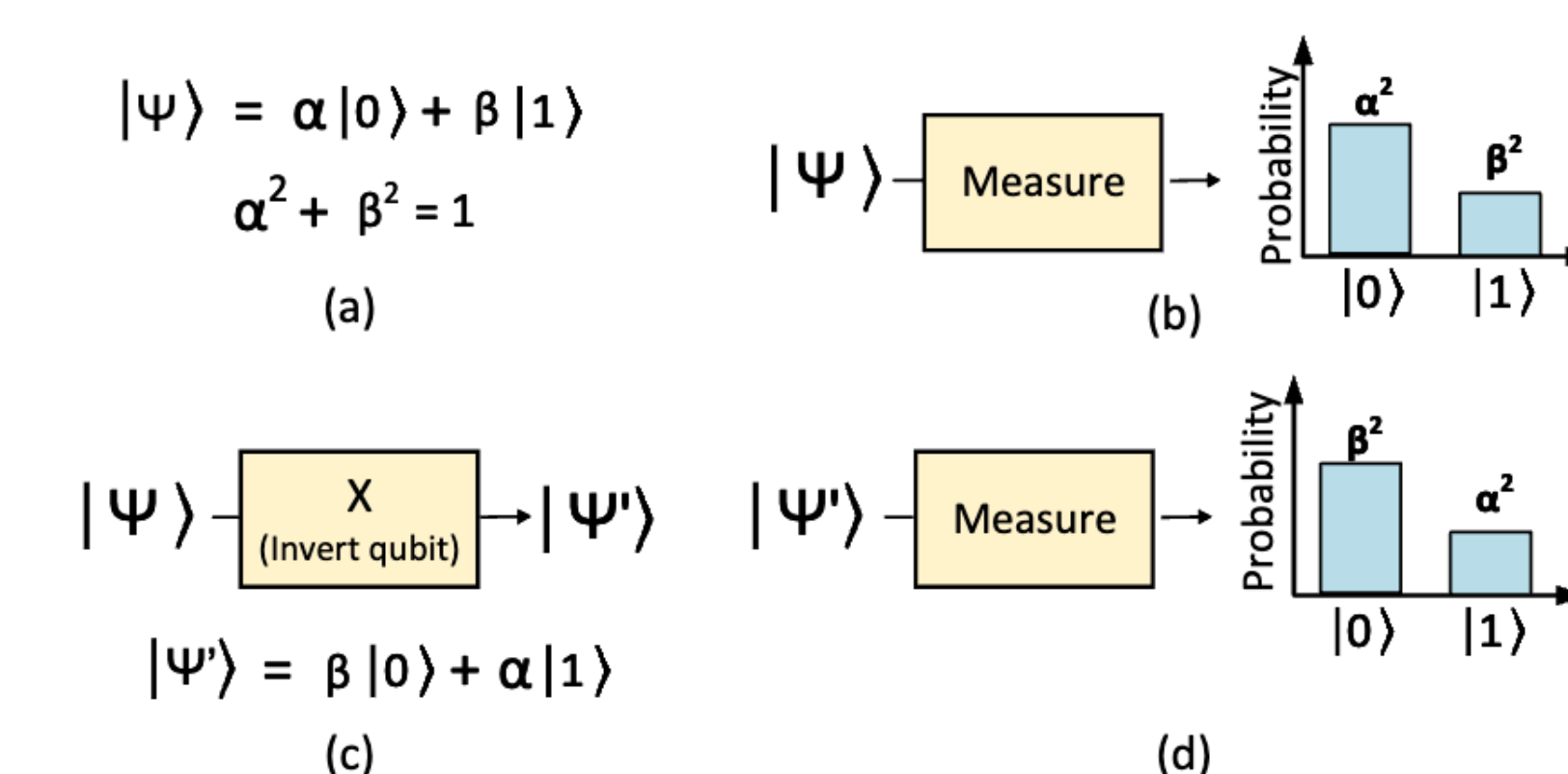
**PROBLEM:** Qubits tend to relax to their lower energy state (0), causing measurement bias that favors detecting 0 over 1



**Figure 2: On five qubit ibmqx4, probability of successfully measuring a state (a) All-zero state "00000" (b) All-ones state "11111"**

### MITIGATION STRATEGY

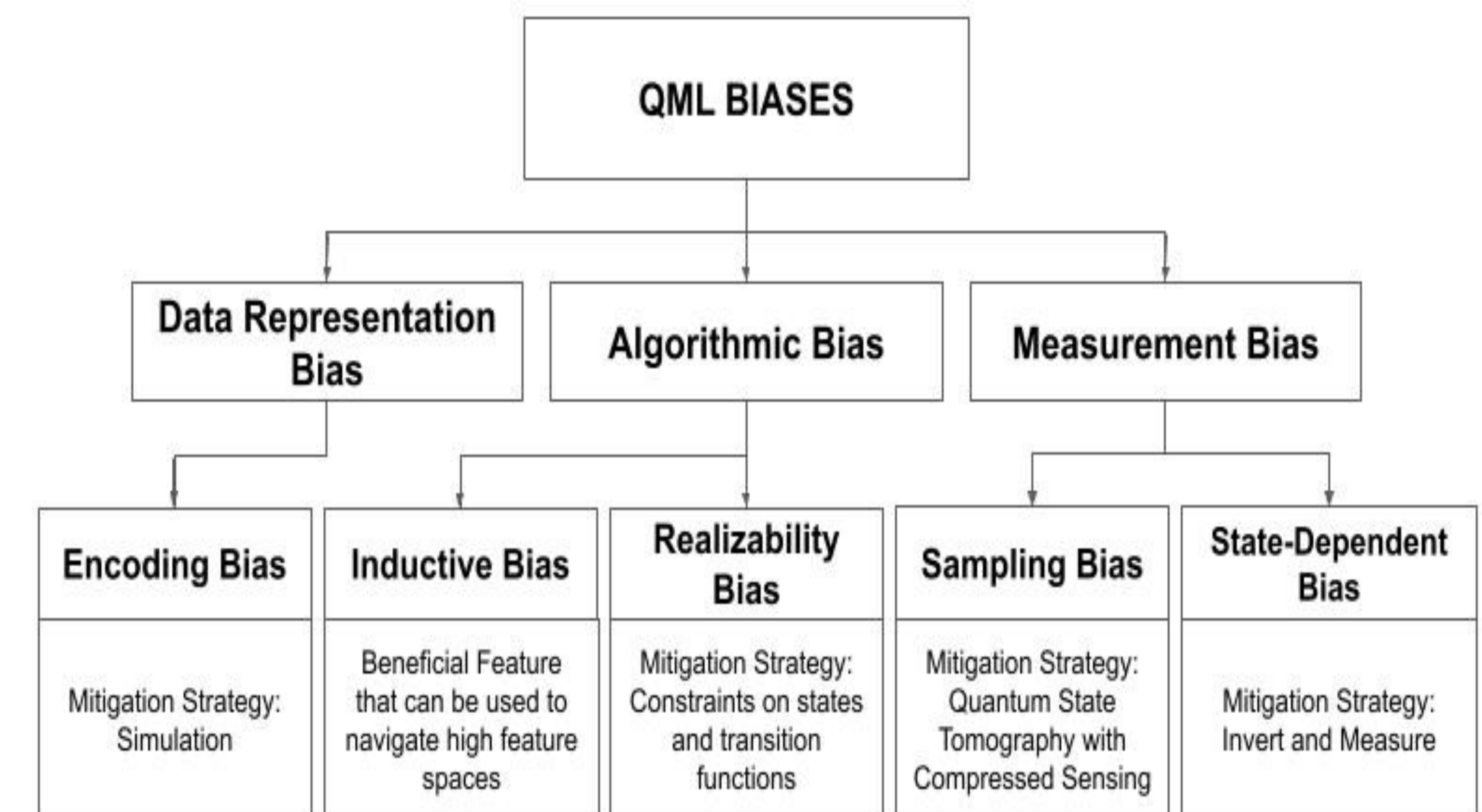
(Adapted from Tannu et. al 2019)



**Figure 3: a) Quantum state is a superposition of  $|0\rangle$  and  $|1\rangle$  (b) Measurement of a qubit is probabilistic (c) X-gate inverts the qubit state (d) Measurement of inverted qubit**

## Overview

This previous examples illustrates some of the critical challenges, but it's just one piece of a larger puzzle. We have identified other key biases that can arise through features of the data, algorithms, or measurements and we consider each of them.



**Figure 4: Overview of biases unique to QML**

## Takeaways

- QML systems are subject to unique biases arising from quantum properties. These biases can significantly impact the fairness, reliability, and performance of QML models.
- Understanding and addressing these biases is crucial as QML applications expand into high-stakes domains.
- Current mitigation strategies exist, but more research is needed to fully address these challenges.