

# Near-optimal Quantum Algorithms for Multivariate Mean Estimation

## Problem setup

1. A probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ ,
2. A  $d$ -dimensional random variable  $X : \Omega \rightarrow \mathbb{R}^d$ .

Properties:

- **Mean:**  $\mu = \mathbb{E}[X] = \sum_{\omega \in \Omega} \mathbb{P}(\omega) X(\omega) \in \mathbb{R}^d$ .
- **Covariance matrix:**  $\Sigma = \begin{bmatrix} \text{Var}[X_1] & \text{Cov}[X_1, X_2] & \cdots & \text{Cov}[X_1, X_d] \\ \text{Cov}[X_1, X_2] & \text{Var}[X_2] & \cdots & \text{Cov}[X_2, X_d] \\ \vdots & \vdots & \ddots & \vdots \\ \text{Cov}[X_1, X_d] & \text{Cov}[X_2, X_d] & \cdots & \text{Var}[X_d] \end{bmatrix}$ .
- **Finite trace:**  $\text{Tr}[\Sigma] = \sum_{j=1}^d \text{Var}[X_j] < \infty$ .

## Access models

**Classical access model:**

1. Obtain outcome  $\omega \sim \mathbb{P}$ .
2. Evaluate function  $\omega \mapsto X(\omega)$ .

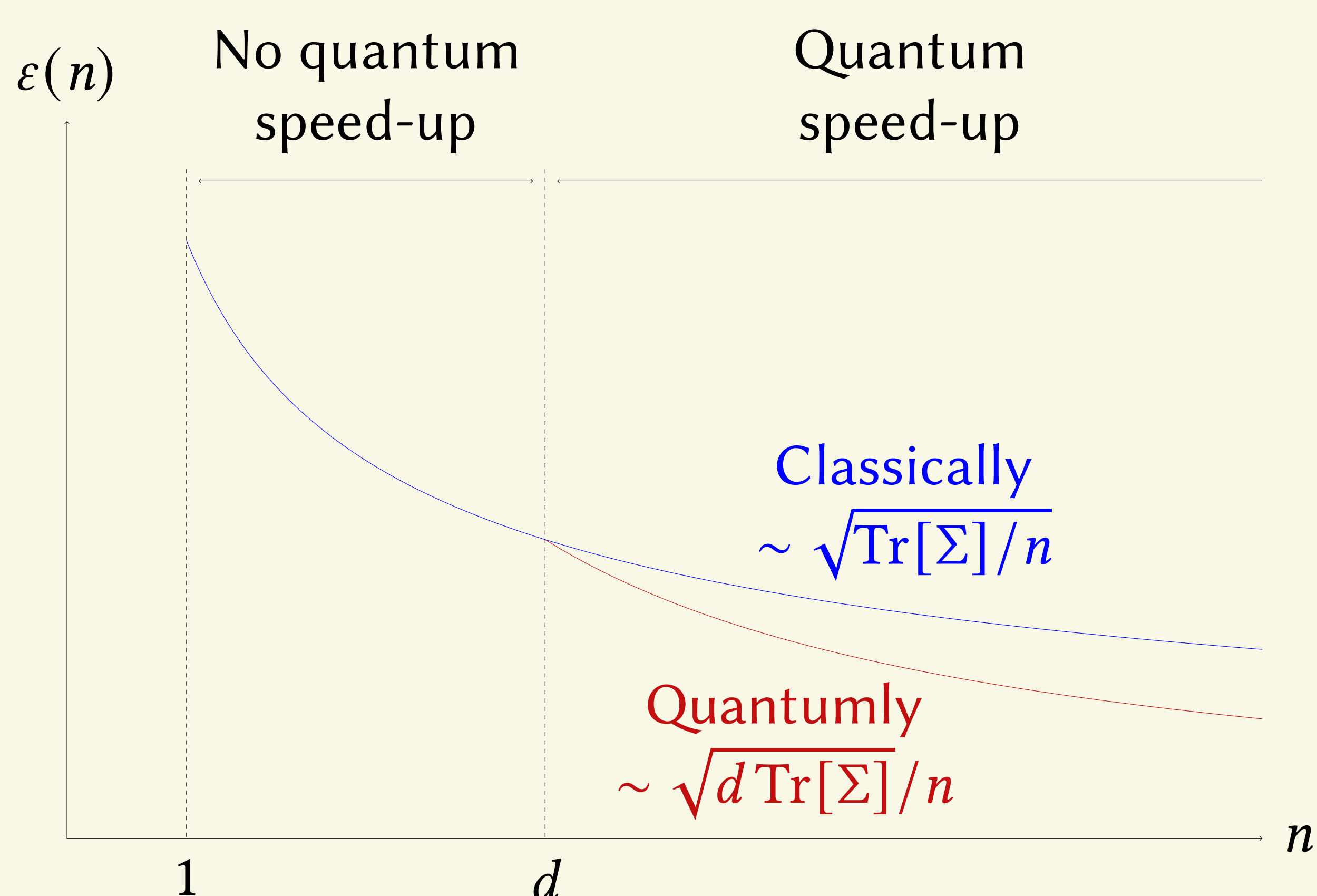
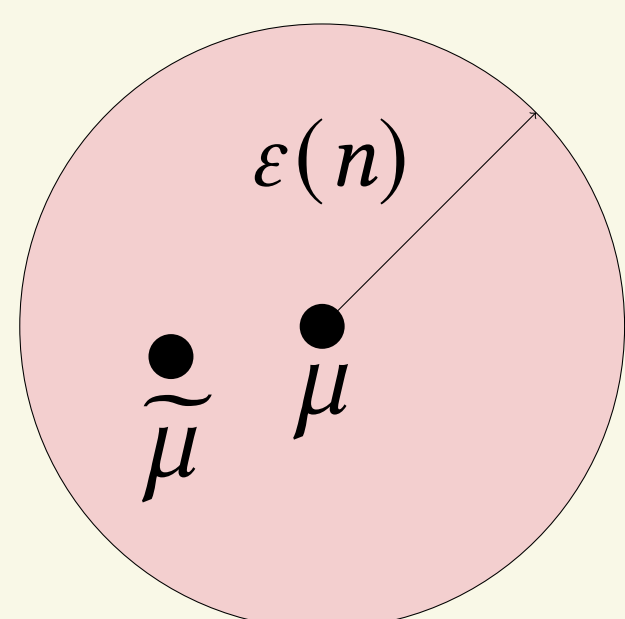
**Quantum access model:**

1. Distribution oracle:  $|0\rangle \mapsto \sum_{\omega \in \Omega} \sqrt{\mathbb{P}(\omega)} |\omega\rangle$ .
2. Evaluation oracle:  $|\omega\rangle |0\rangle \mapsto |\omega\rangle |X(\omega)_1\rangle \otimes \cdots \otimes |X(\omega)_d\rangle$ .

↳ Calls to these routines are **samples**.

## Results

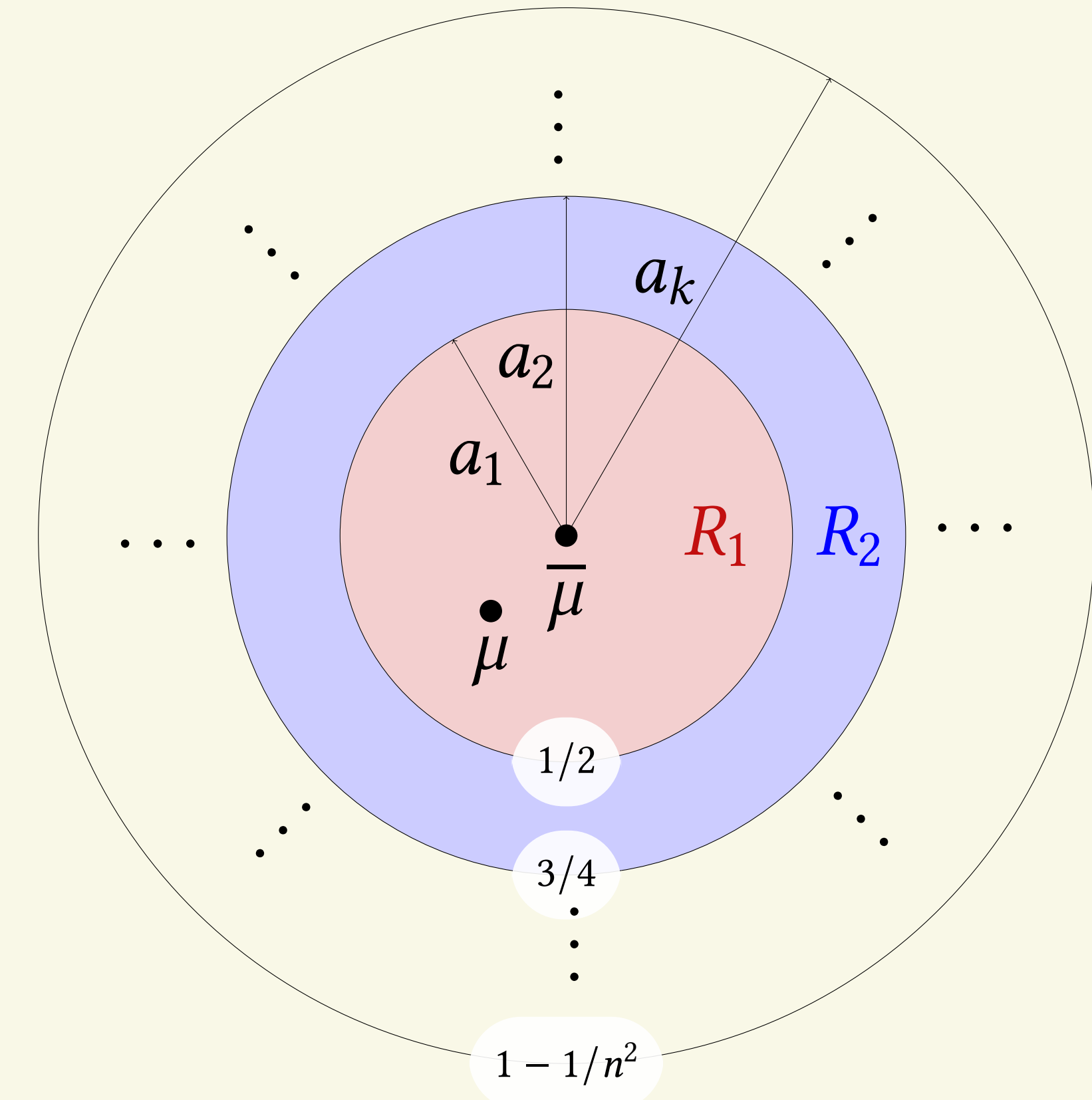
**Goal:** Construct an estimator  $\tilde{\mu}$ , using  $n$  samples, s.t.  $\mathbb{P}[\|\mu - \tilde{\mu}\|_2 \leq \varepsilon(n)] \geq \frac{2}{3}$ .



**Crucial observation:** quantum speed-up only when  $n \geq d$ .

## Quantum estimator outline

1. **Get a crude estimate:**  $\bar{\mu}$  s.t.  $\|\mu - \bar{\mu}\|_2 \leq \sqrt{\text{Tr}[\Sigma]}$ .
2. **Get an idea of the spread:** Estimate quantiles  $a_\ell$  s.t.  $\mathbb{P}[\|X - \bar{\mu}\|_2 \geq a_\ell] \approx \frac{1}{2^\ell}$ , for  $\ell \in \{1, \dots, 2 \log(n)\}$ .
3. **Estimate truncated mean on the rings:**  $\mathbb{E}[X] \approx \sum_{\ell=1}^k \mathbb{E}[X^{(\ell)}]$ .

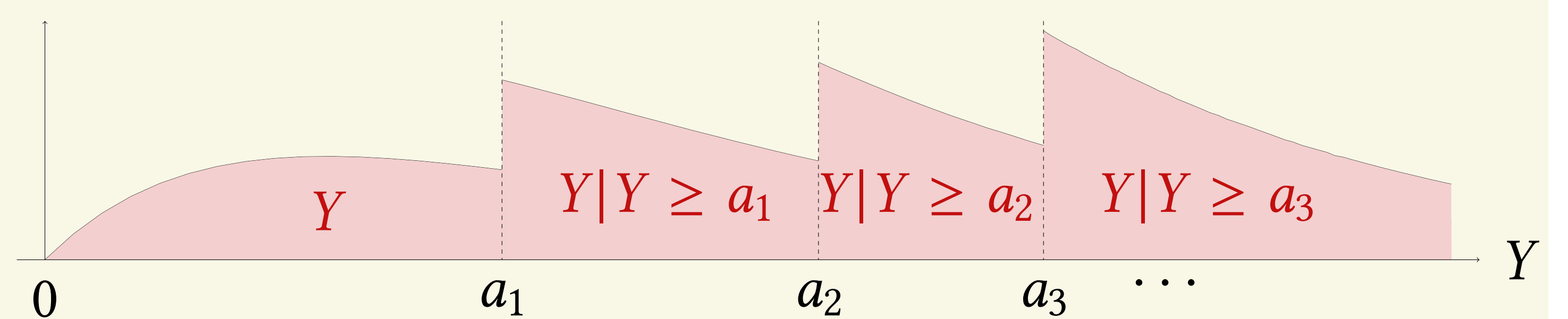


## Quantile estimation

Define the univariate random variable  $Y = \|X - \bar{\mu}\|_2$ .

**Observation:**  $a_\ell \approx$  median of  $\tilde{O}(1)$  samples from  $Y|Y \geq a_{\ell-1}$ .

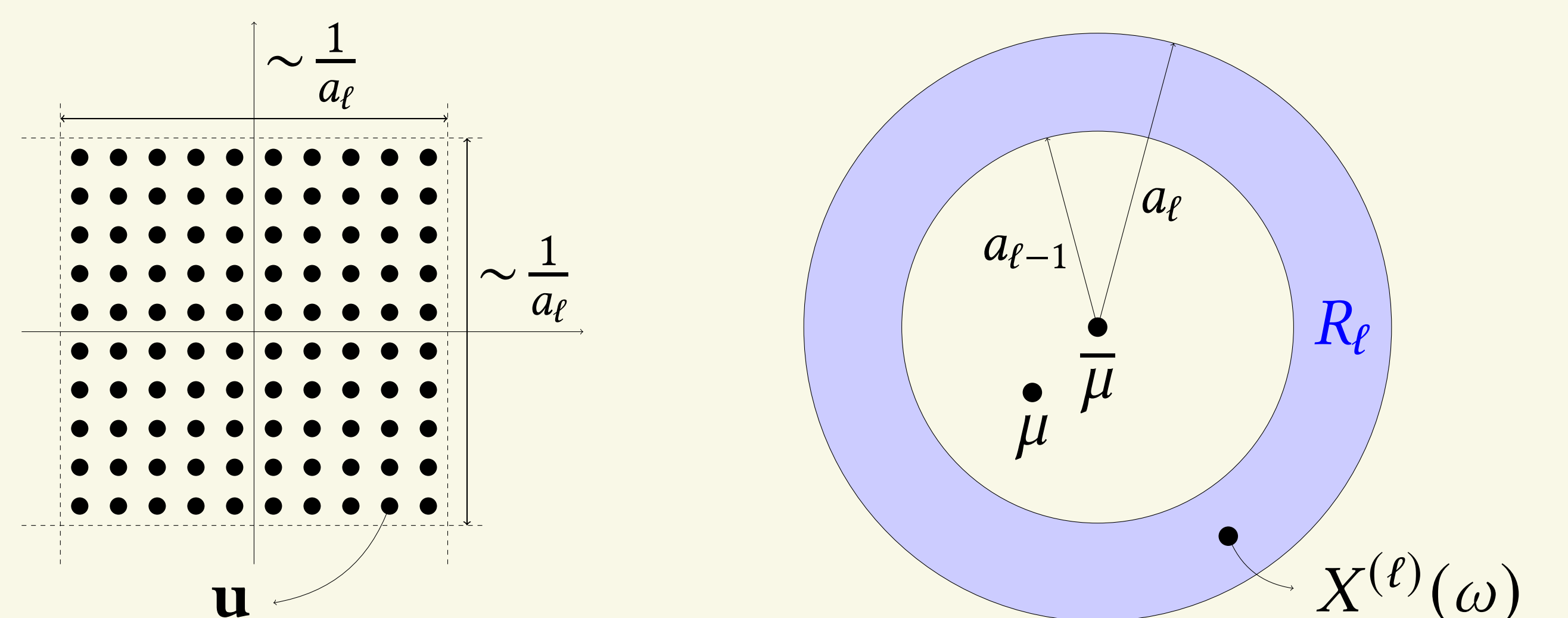
Rejection sampling via **amplitude amplification** requires  $\tilde{O}(1/\sqrt{\mathbb{P}[Y \geq a_{\ell-1}]}) = \tilde{O}(\sqrt{2}^\ell)$  samples from  $Y$ .



## Truncated mean estimation on the rings

1. **Amplitude amplification on the  $R_\ell$  ring:**  $\tilde{O}(\sqrt{2}^\ell)$  samples.
2. **Phase encoding techniques:**  $\tilde{O}(n/\sqrt{2}^\ell)$  calls to **1**.

Directional mean oracle:  $|\mathbf{u}\rangle \mapsto e^{in2^{\ell/2} \cdot \mathbf{u}^T \mathbb{E}[X^{(\ell)}]} |\mathbf{u}\rangle$ .



3. **Bernstein-Vazirani over the reals:**  $\tilde{O}(1)$  calls to **2**.

Apply  $\text{QFT}^{-1}$  on  $\sum_{\mathbf{u}} e^{in2^{\ell/2} \cdot \mathbf{u}^T \mathbb{E}[X^{(\ell)}]} |\mathbf{u}\rangle$  and measure.

$$\|\tilde{\mu}^{(\ell)} - \mathbb{E}[X^{(\ell)}]\| = \tilde{O}(\sqrt{d \cdot a_\ell \cdot \mathbb{E}\|X^{(\ell)}\|/n}) = \tilde{O}(\sqrt{d \cdot \text{Tr}[\Sigma]/n})$$