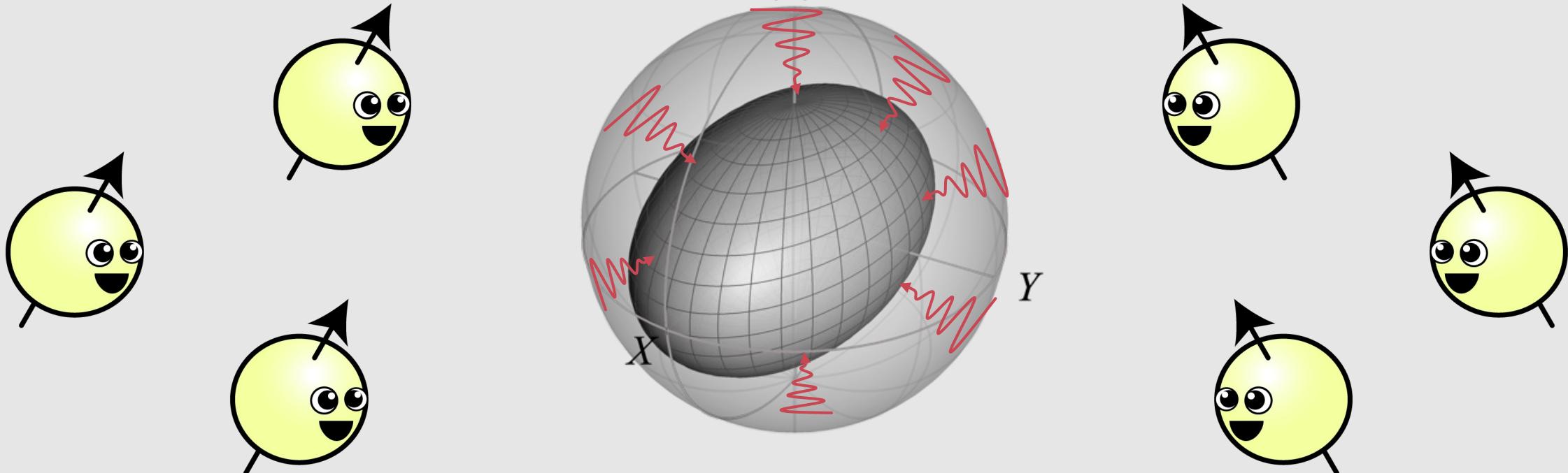


# Introduction to Quantum Noise

IBM Quantum

*Qiskit Global Summer School: Quantum Simulations*



**Zlatko K. Minev**

IBM Quantum

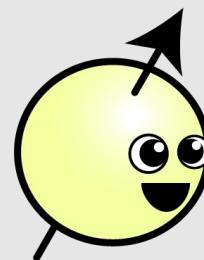


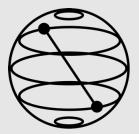
@zlatko\_minev



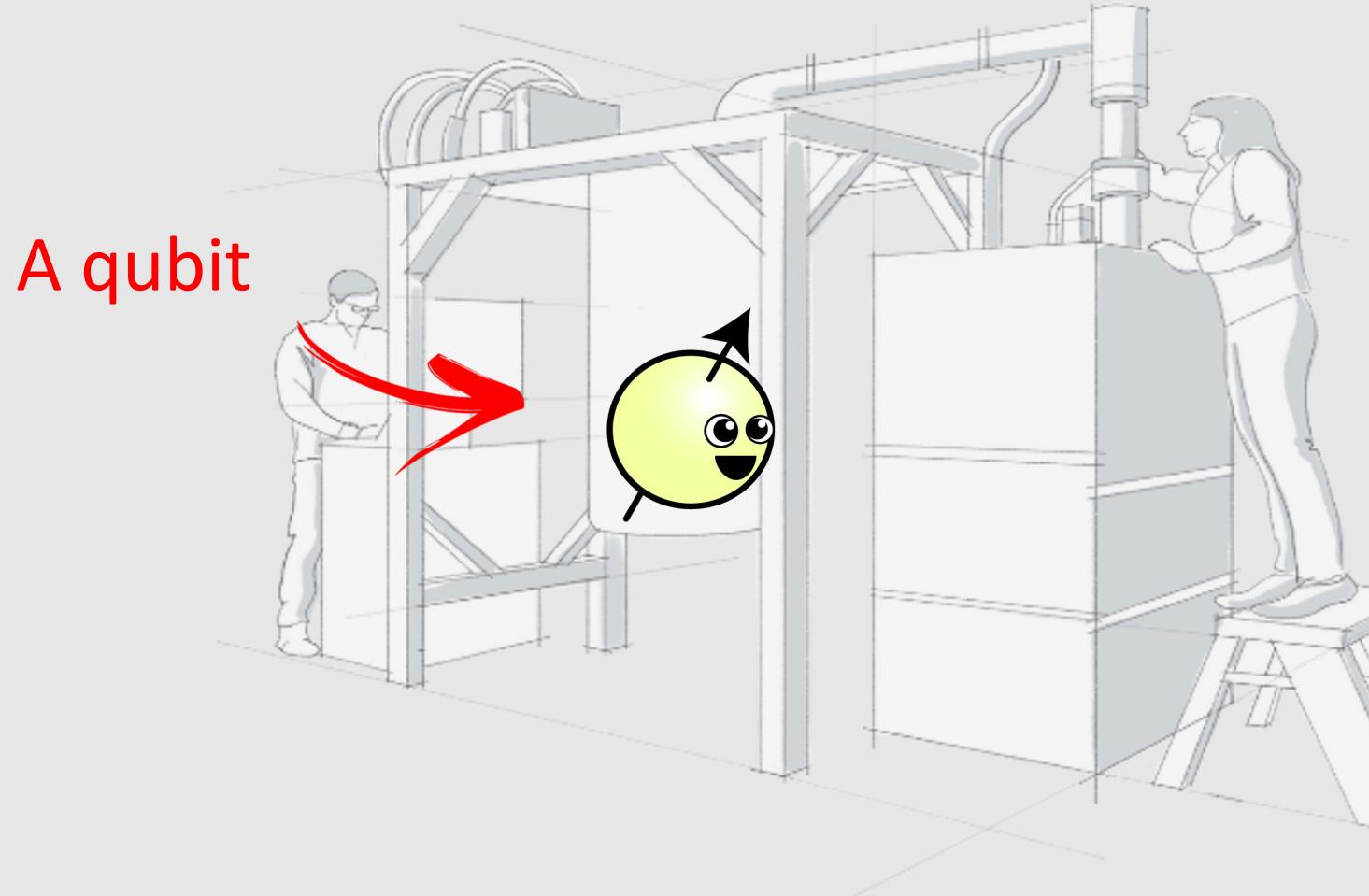
[zlatko-minev.com](http://zlatko-minev.com)

What do I need to know before  
I run quantum simulation on a  
real, noisy quantum processor?



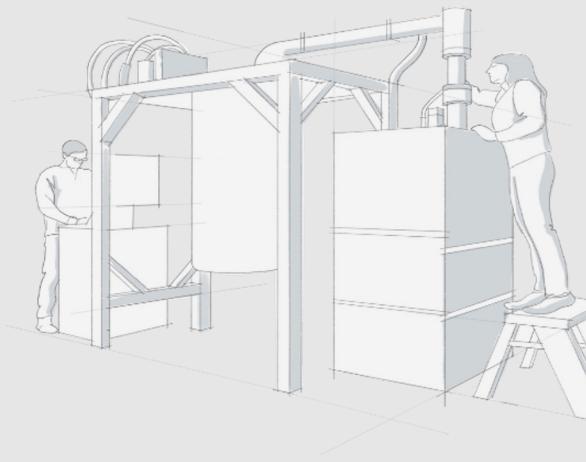


# Chapter 1: Hello World!

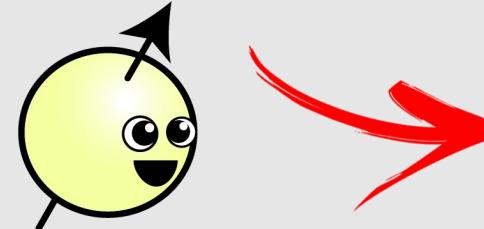




# Hello World! building blocks



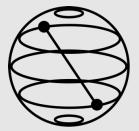
A qubit



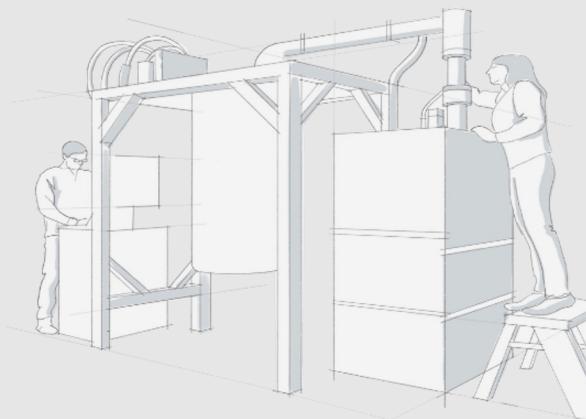
$|1\rangle$

$|0\rangle$

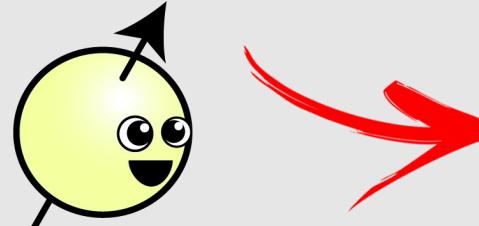
Computational  
basis states



# Hello World! building blocks

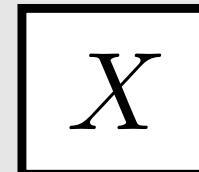


A qubit

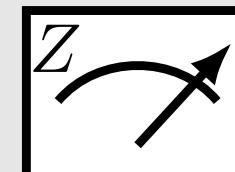
 $|1\rangle$  $|0\rangle$ 

Computational  
basis states

Operations: qubit gate



Measurements: qubit observable



refresher:

$$X |0\rangle = |1\rangle$$

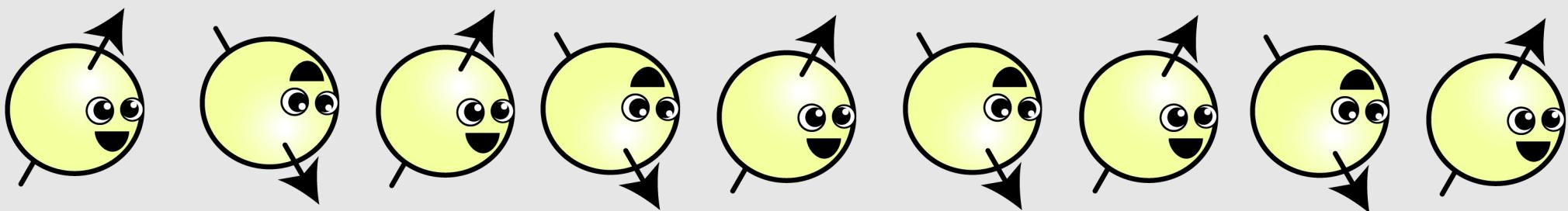
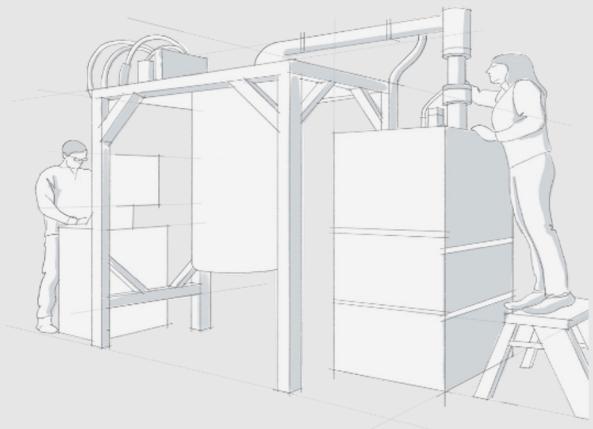
$$X |1\rangle = |0\rangle$$

$$Z |0\rangle = +1 |0\rangle$$

$$Z |1\rangle = -1 |1\rangle$$



# Hello World! Even-odd algo: qubit flipper



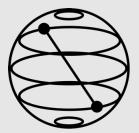
refresher:

$$X |0\rangle = |1\rangle$$

$$X |1\rangle = |0\rangle$$

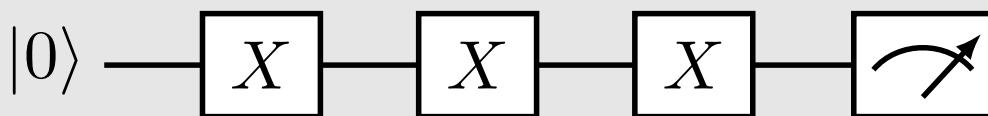
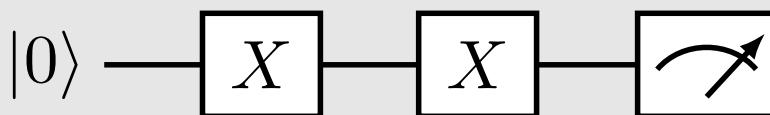
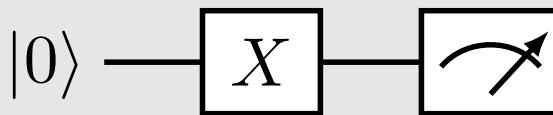
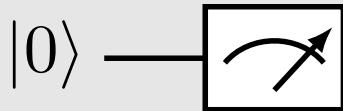
$$Z |0\rangle = +1 |0\rangle$$

$$Z |1\rangle = -1 |1\rangle$$



# Hello World! qubit flipper quantum circuits

depth



⋮

refresher:

$$X |0\rangle = |1\rangle$$

$$X |1\rangle = |0\rangle$$

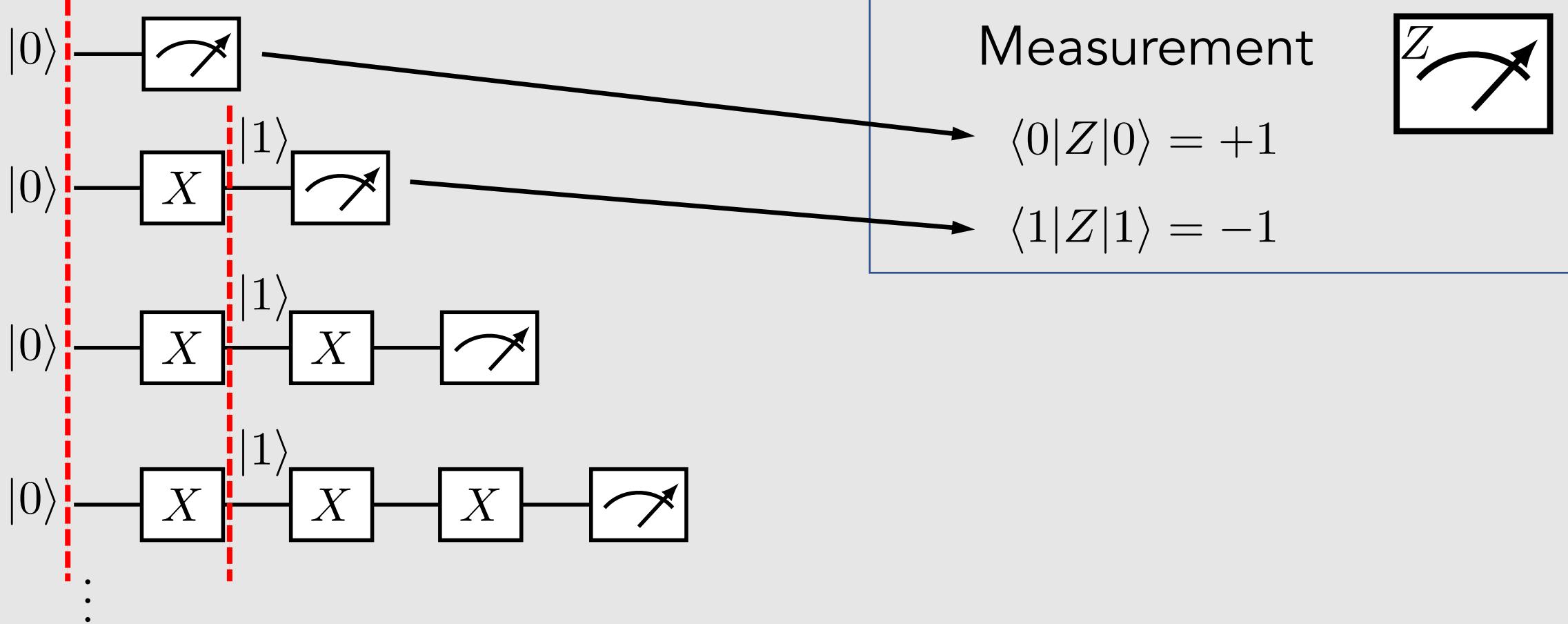
$$Z |0\rangle = +1 |0\rangle$$

$$Z |1\rangle = -1 |1\rangle$$



# Hello World! “debugger” step through

depth



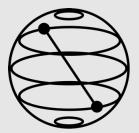
refresher:

$$X |0\rangle = |1\rangle$$

$$X |1\rangle = |0\rangle$$

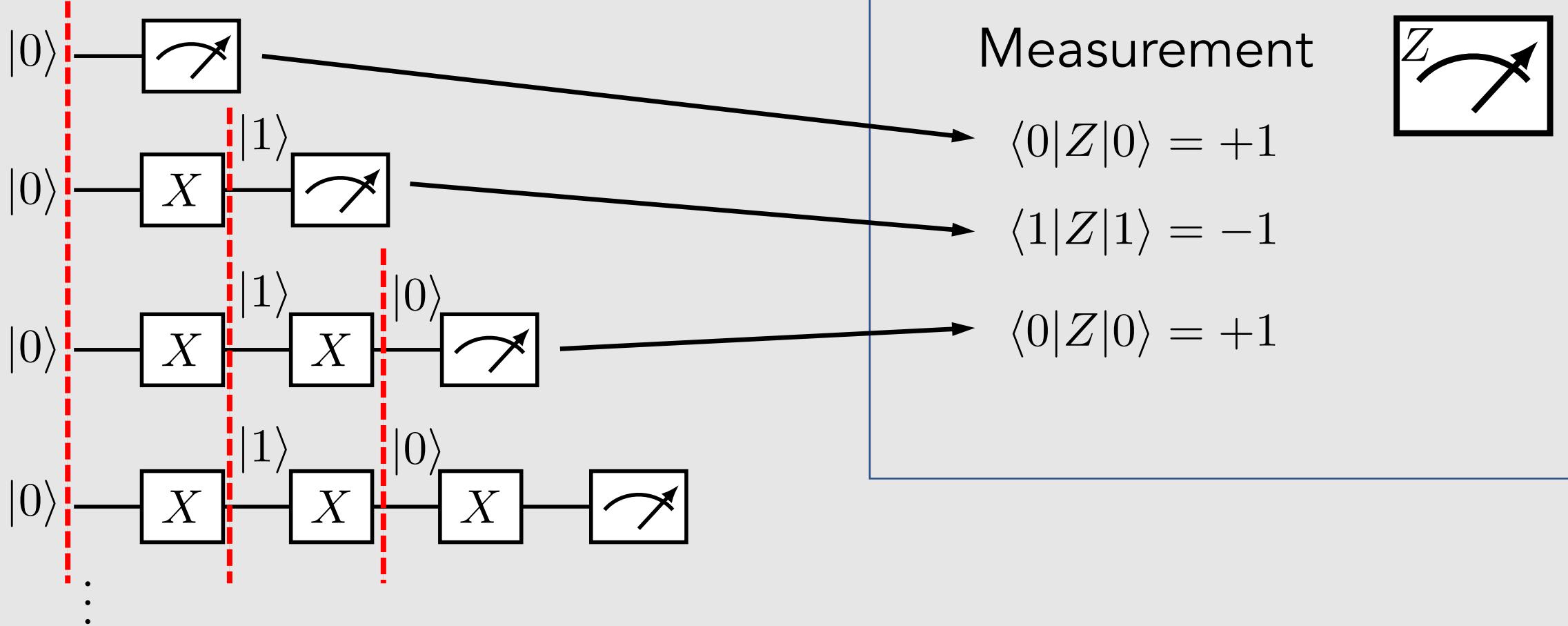
$$Z |0\rangle = +1 |0\rangle$$

$$Z |1\rangle = -1 |1\rangle$$



# Hello World! “debugger” step through

depth



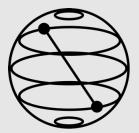
refresher:

$$X |0\rangle = |1\rangle$$

$$X |1\rangle = |0\rangle$$

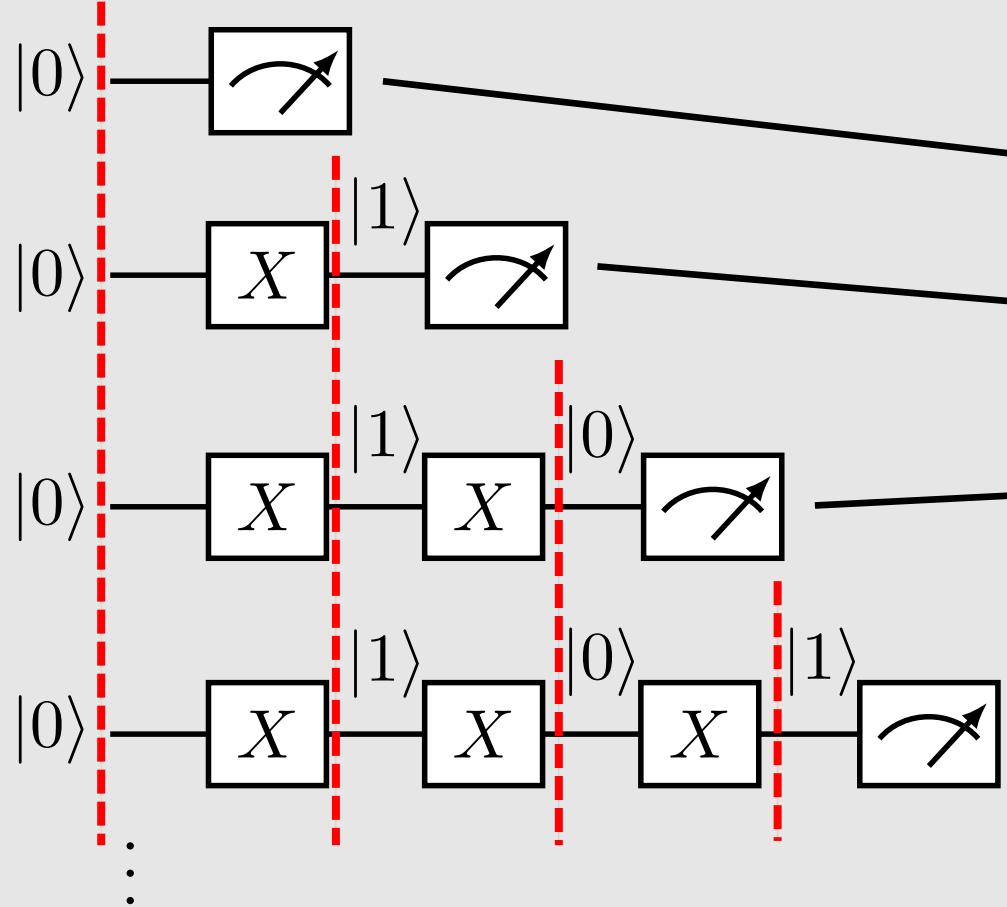
$$Z |0\rangle = +1 |0\rangle$$

$$Z |1\rangle = -1 |1\rangle$$

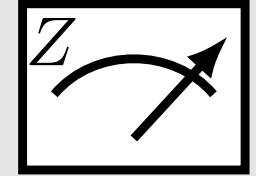


# Hello World! “debugger” step through

depth



Measurement



$$\langle 0|Z|0\rangle = +1$$

$$\langle 1|Z|1\rangle = -1$$

$$\langle 0|Z|0\rangle = +1$$

$$\langle Z \rangle = (-1)^d$$

where  $d$  is the circuit depth

refresher:

$$X |0\rangle = |1\rangle$$

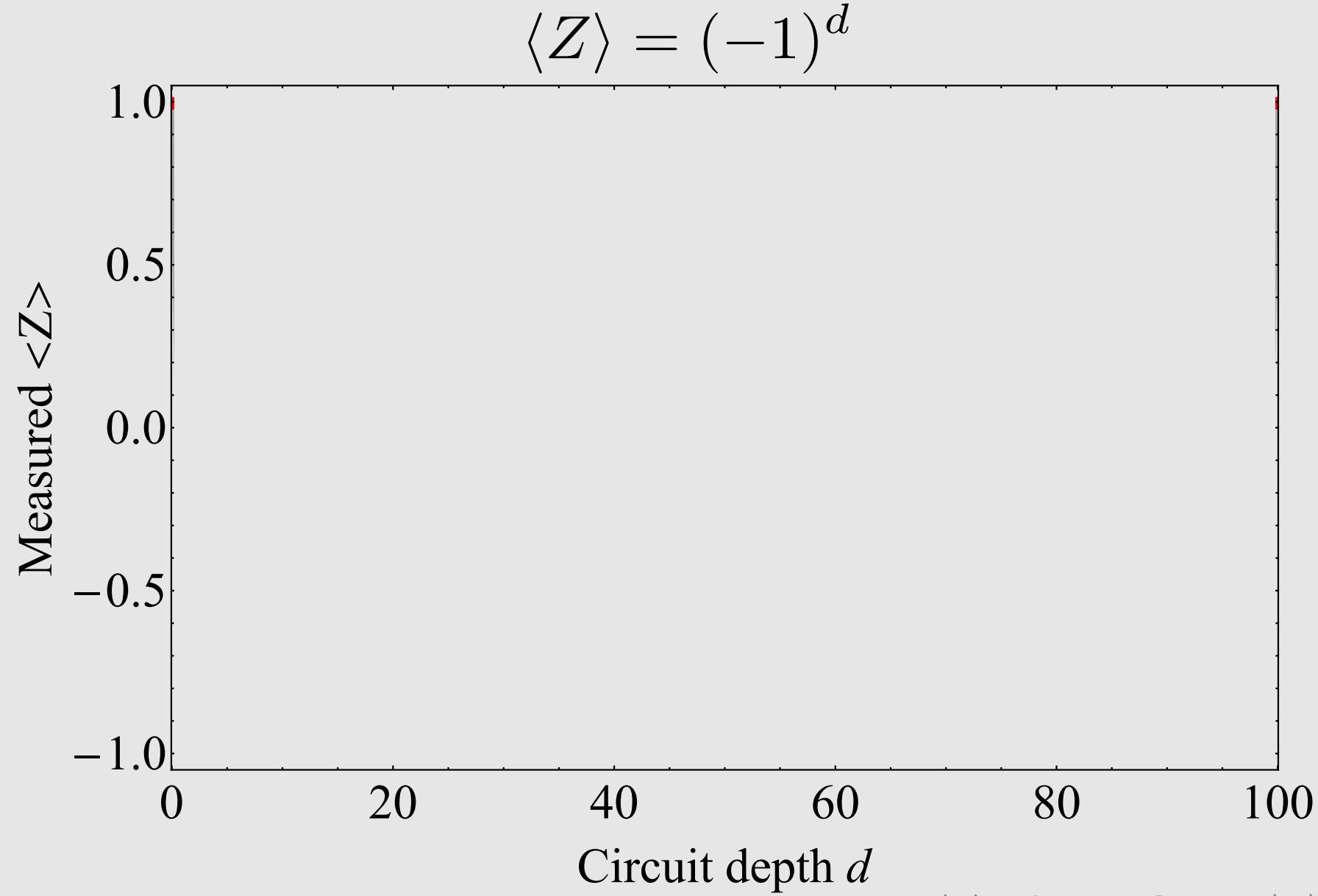
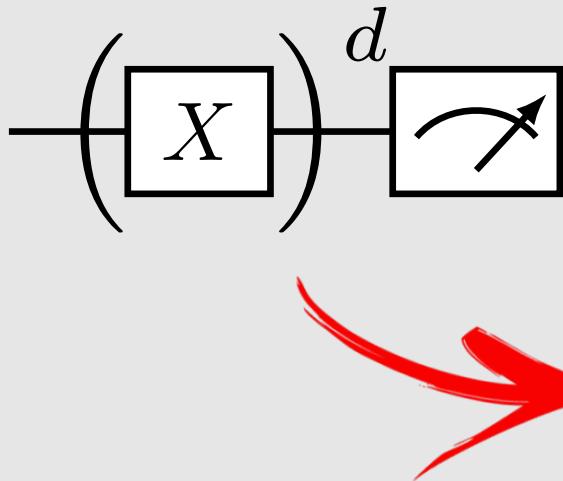
$$X |1\rangle = |0\rangle$$

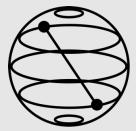
$$Z |0\rangle = +1 |0\rangle$$

$$Z |1\rangle = -1 |1\rangle$$

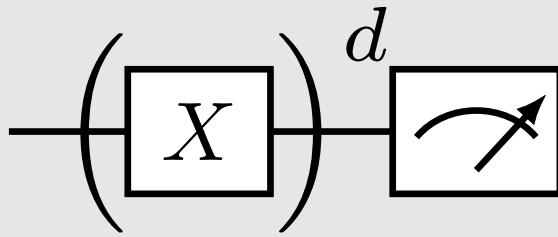


# Hello World! Ideal expectation results

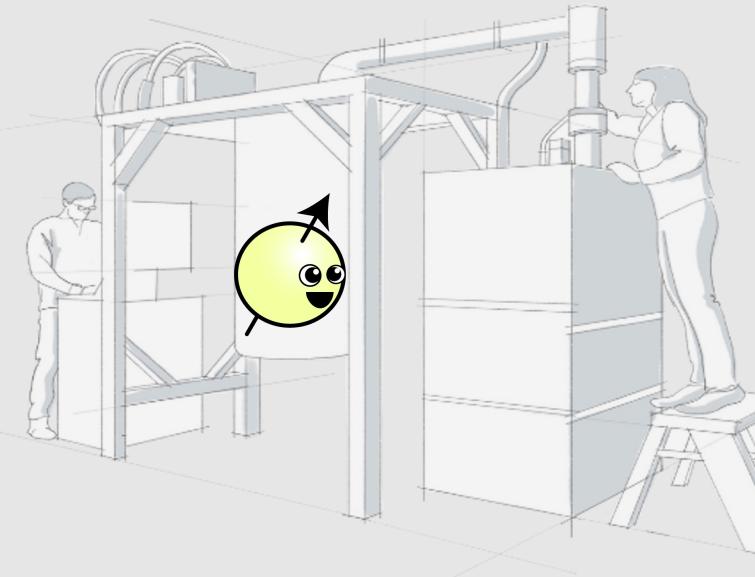




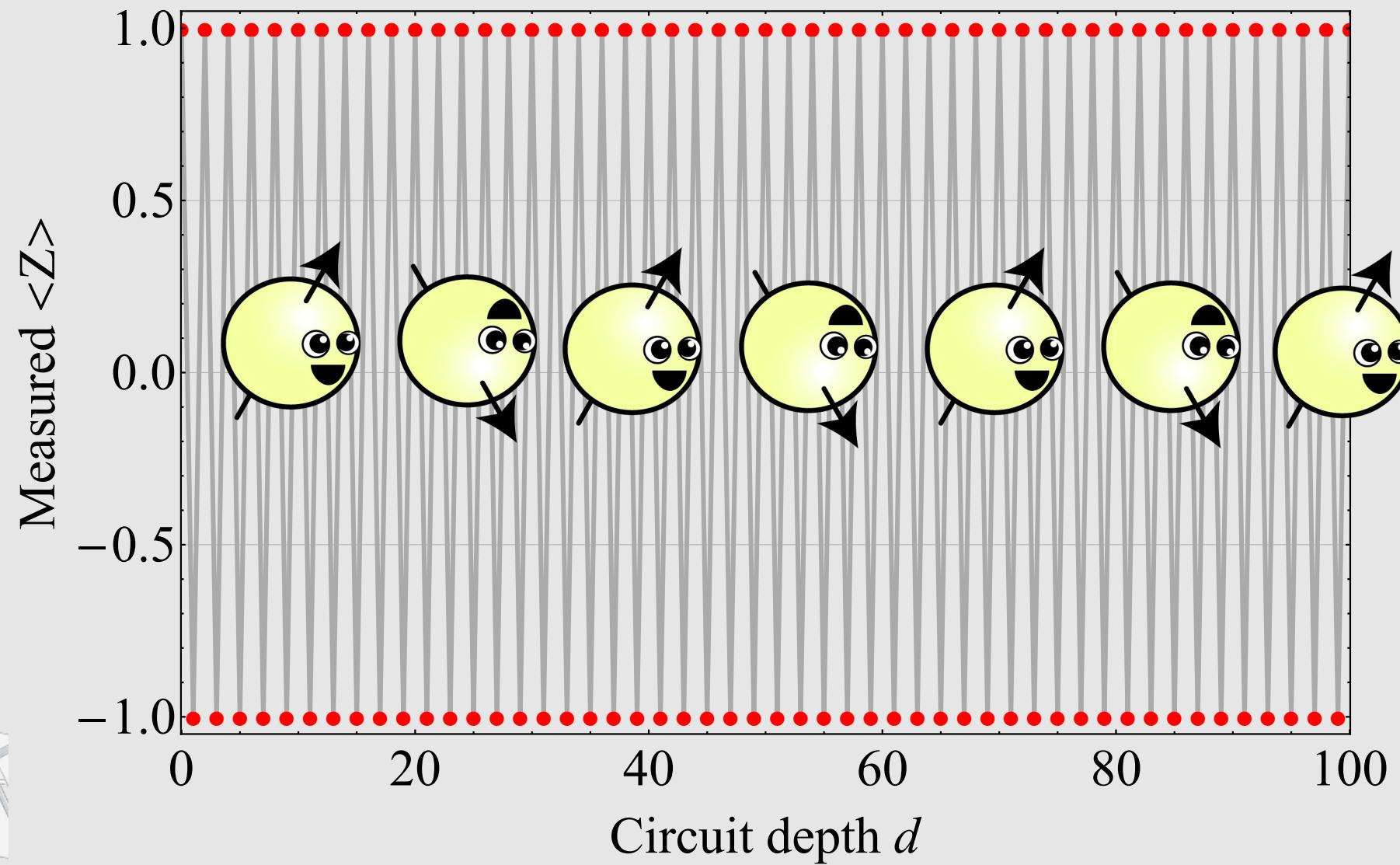
# Hello World! Ideal expectation results

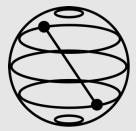


Let's run on a real device!

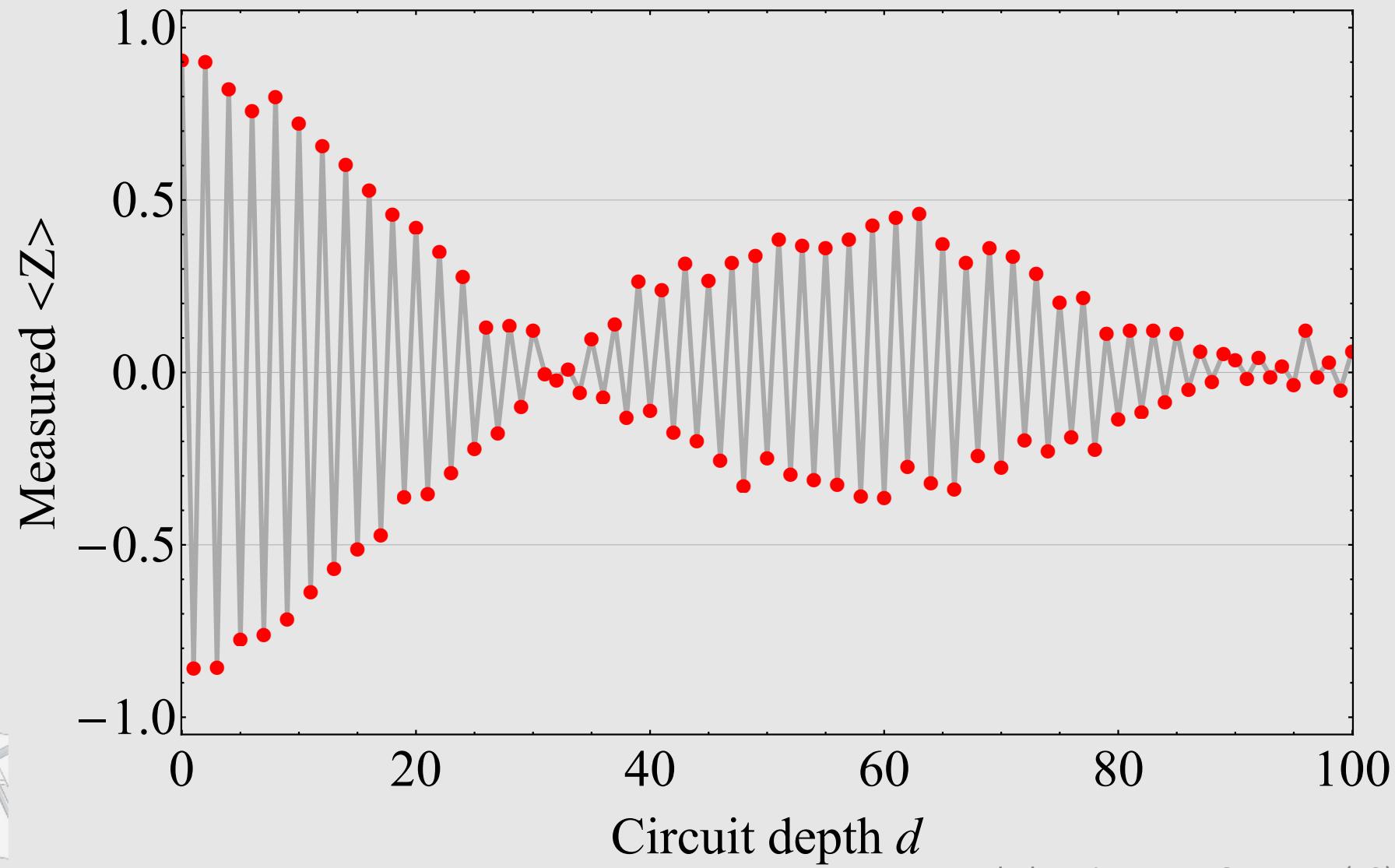


$$\langle Z \rangle = (-1)^d$$



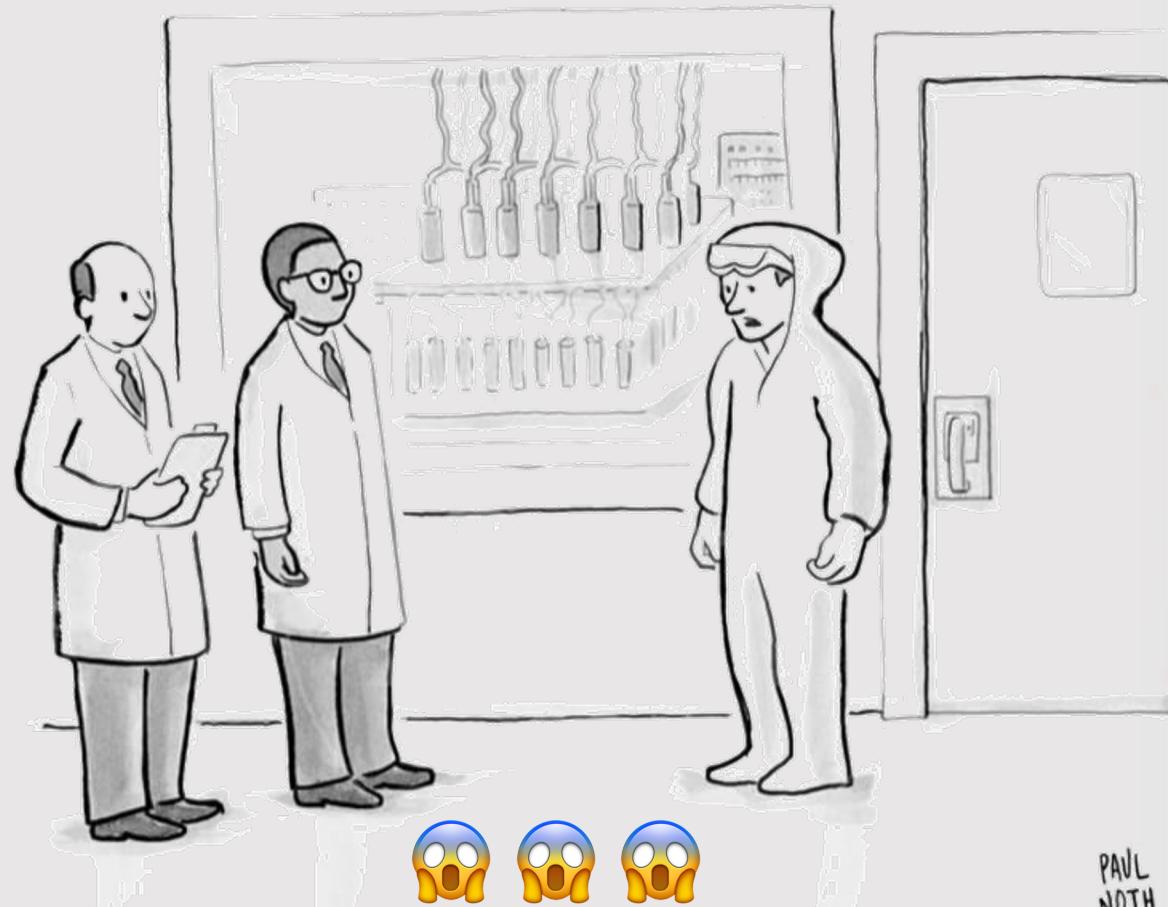


# Hello World! Real expectation results

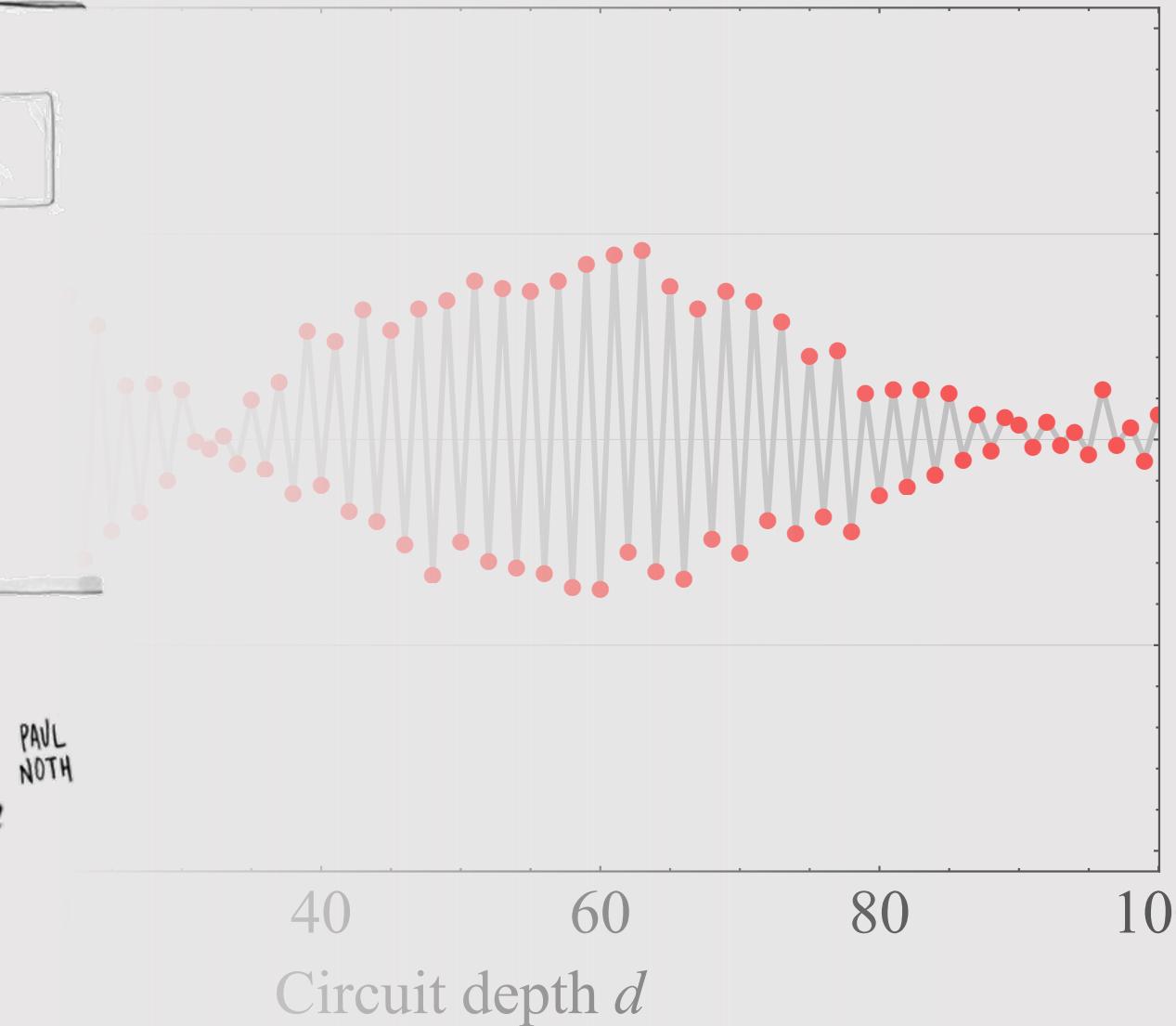


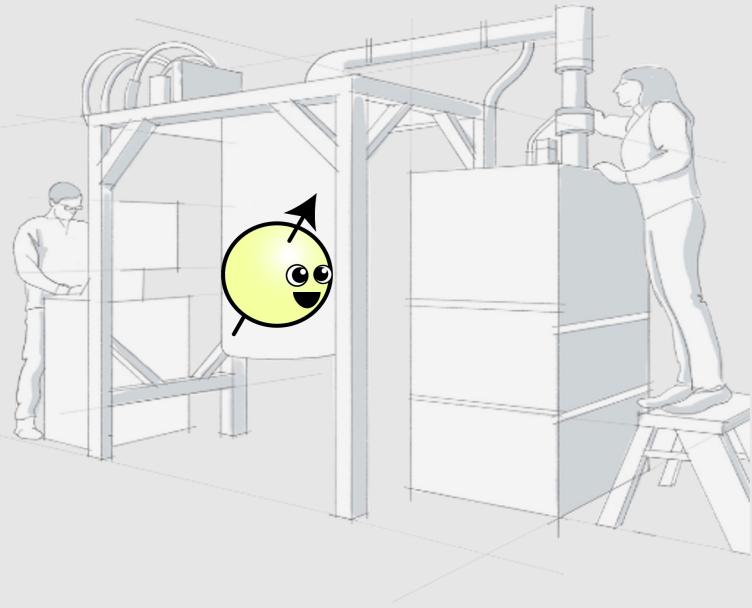


# Real & noisy quantum processors: Why study noise?



*“Well, your quantum computer is broken in every way possible simultaneously.”*



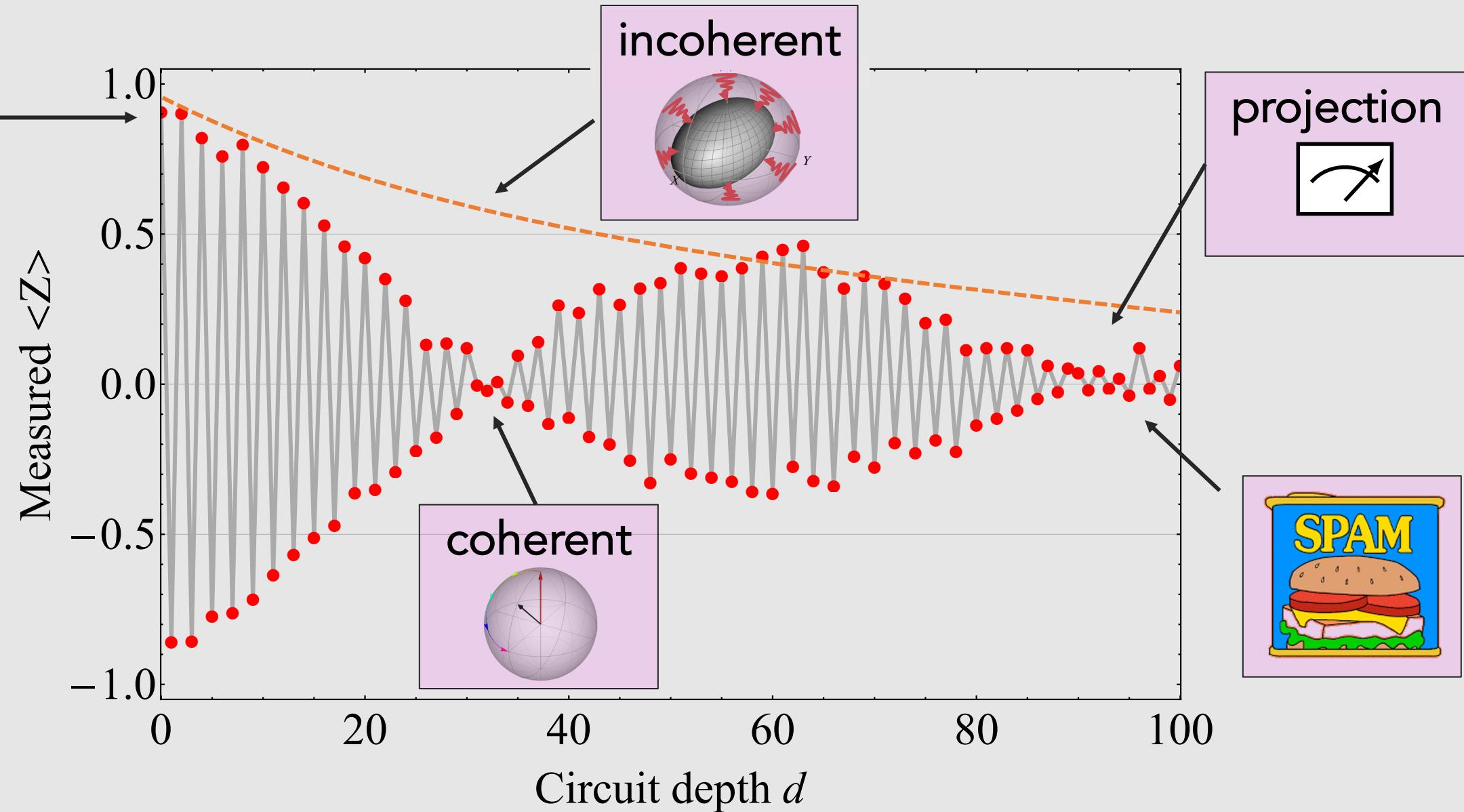


**“Quantum phenomena  
do not occur in a Hilbert space,  
they occur in a laboratory.”**

**Asher Peres**

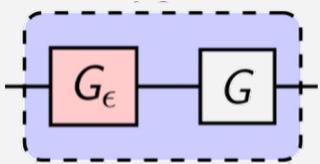
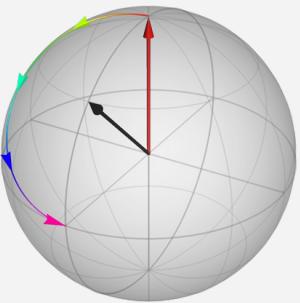


# Elements of 😱 noise

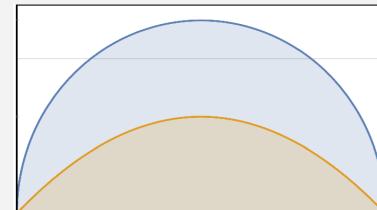


# The road ahead

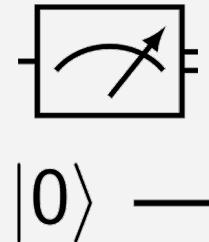
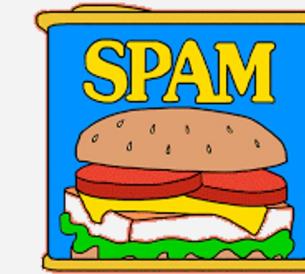
## Coherent noise



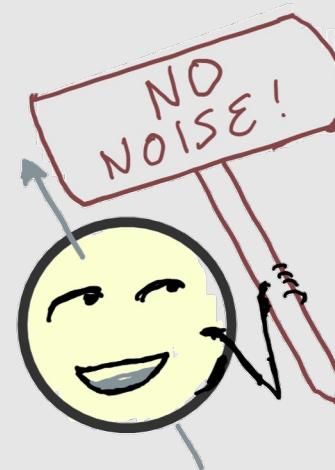
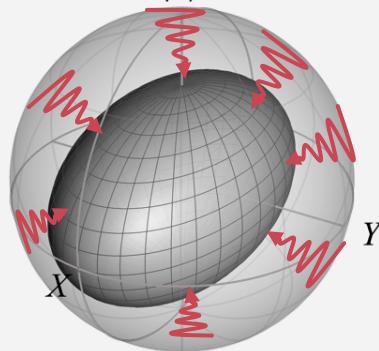
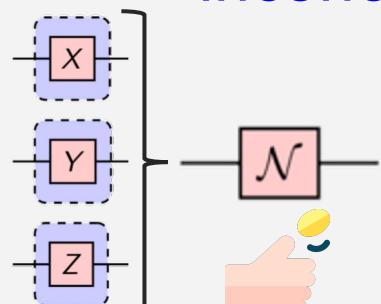
## Measurement in a nutshell Projection noise



## SPAM: Noisy meters



## Incoherent noise



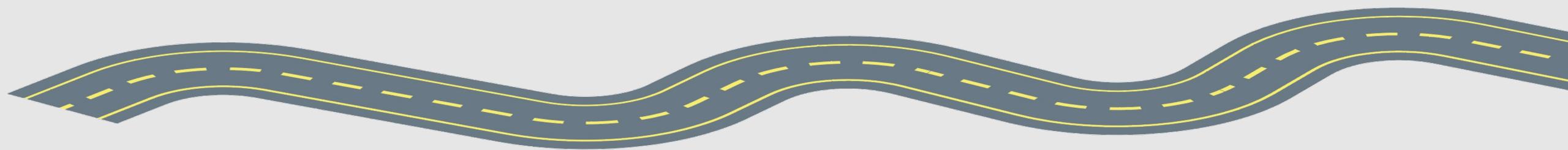
## Bonus content Coherent ZZ State preparation



coin toss: flaticon; spam: make it move;  
road based on: freepik

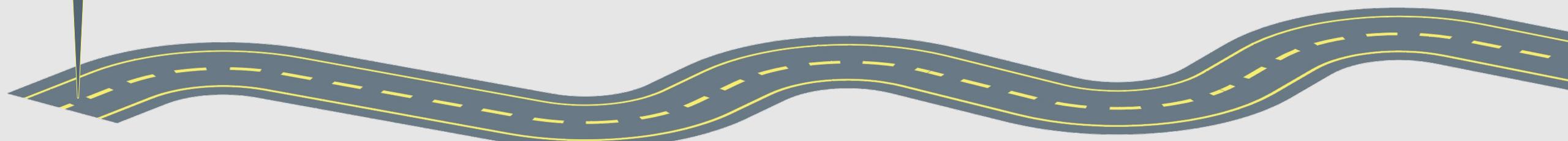
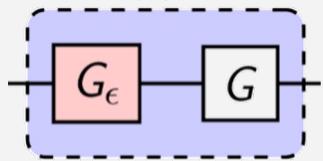
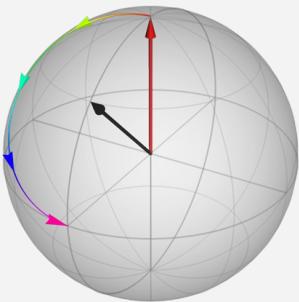
# The destination is the journey

---



# Chapter 2: Coherent noise

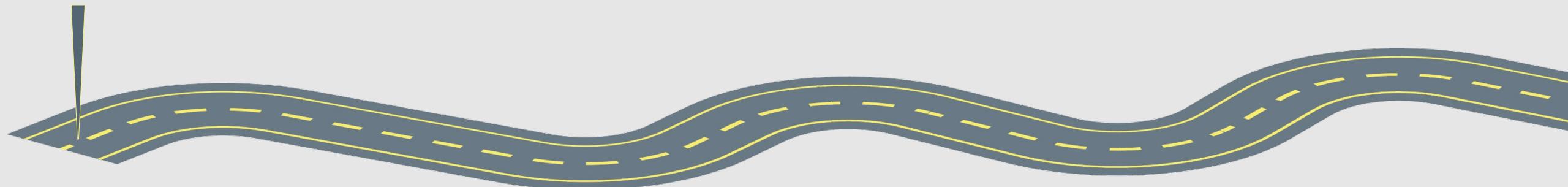
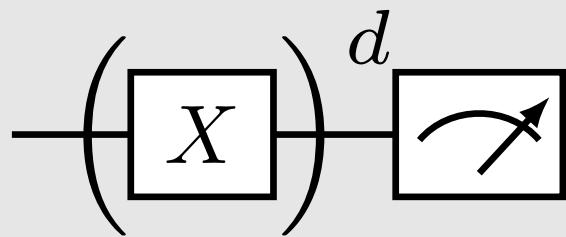
Coherent



road based on: freepik

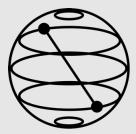
Zlatko Minev, IBM Quantum (19)

# Return to the Hello World example



road based on: freepik

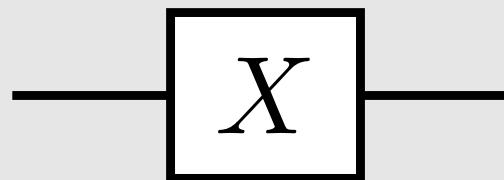
Zlatko Minev, IBM Quantum (20)



# Origin of our X gate: time evolution

$$X = R_X(\pi)$$

(up to global phase)



Refresher:

$$\hat{H} = \frac{\hbar\omega}{2} X$$

$$= \frac{\hbar\omega}{2} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$$U(t) = \exp\left(-it\hat{H}/\hbar\right)$$

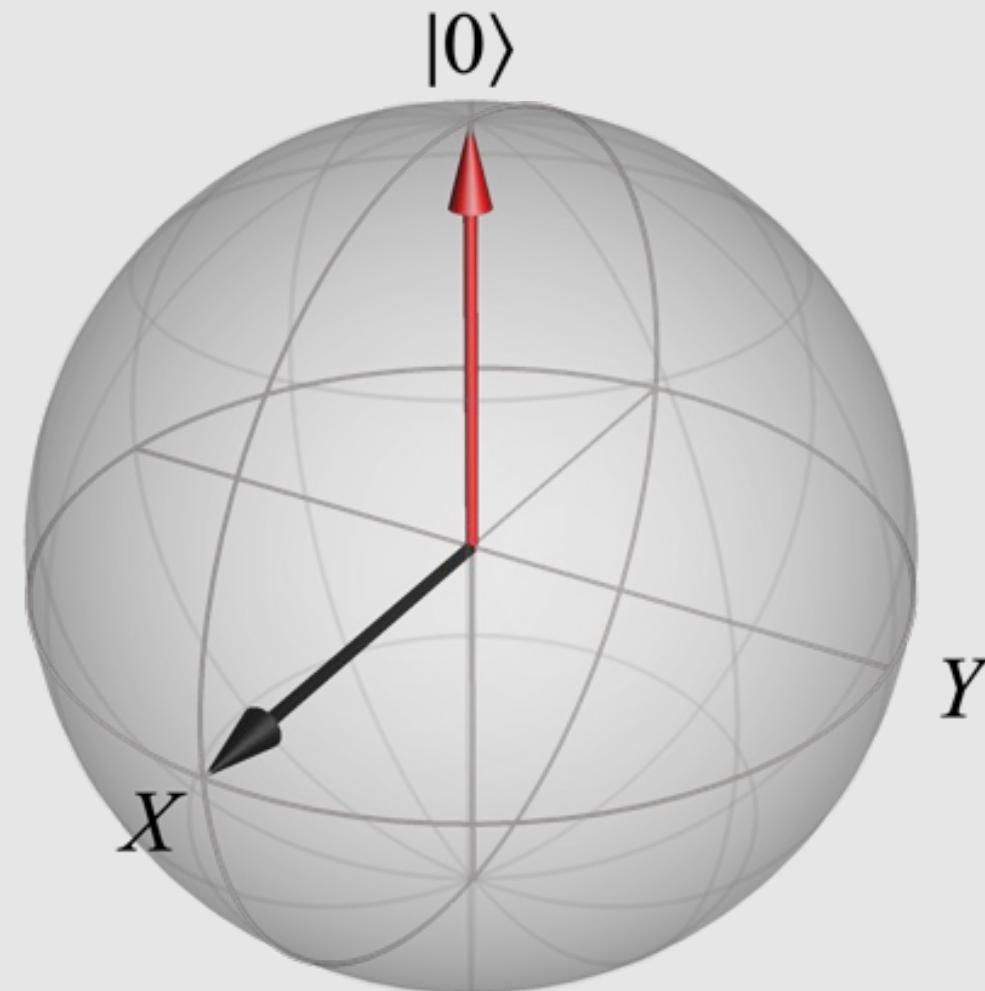
$$\theta := \omega t$$

$$R_X(\theta) = \exp\left(-\frac{i\theta}{2} X\right)$$

$$= \cos(\theta/2)I - i \sin(\theta/2)X$$

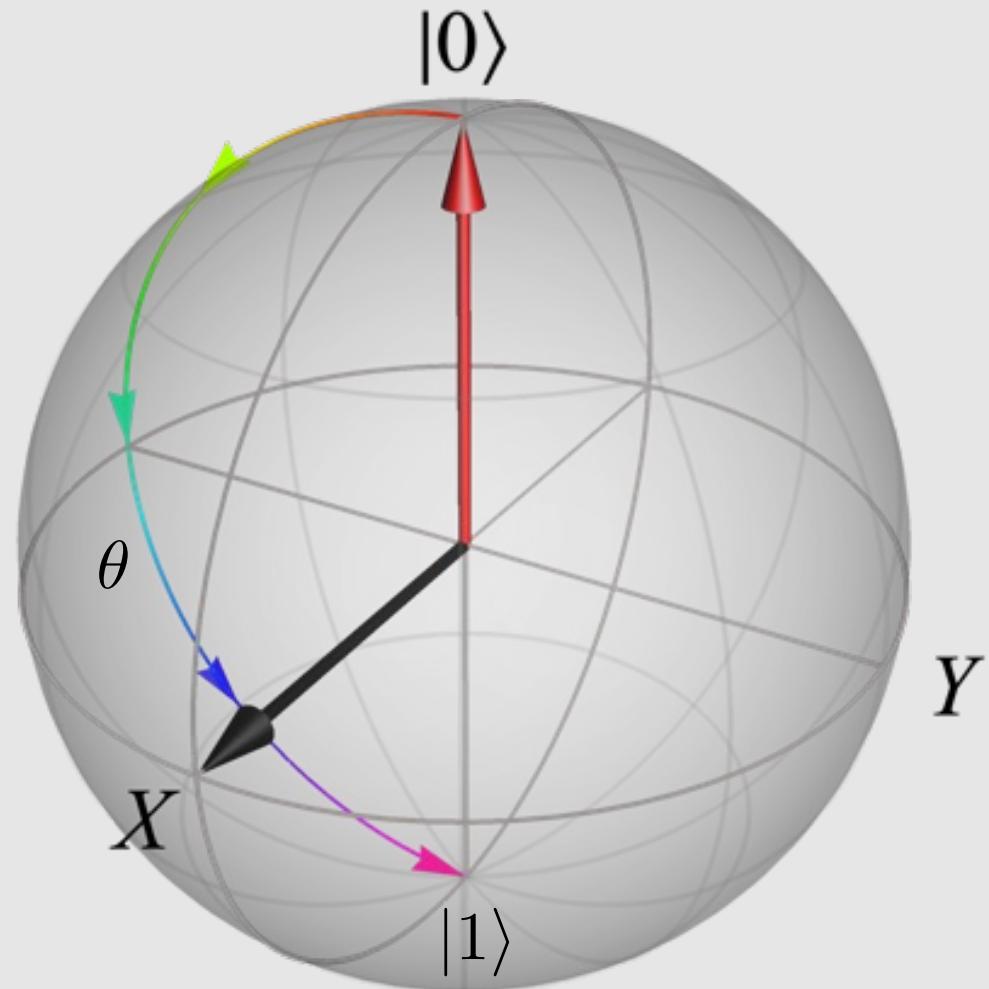
\* We will often drop hats on Paulis  $I, X, Y, Z$

# Visualize: Bloch sphere



# Visualize: Evolution on the Bloch sphere

$$R_X(\theta) = \exp\left(-\frac{i\theta}{2}X\right)$$

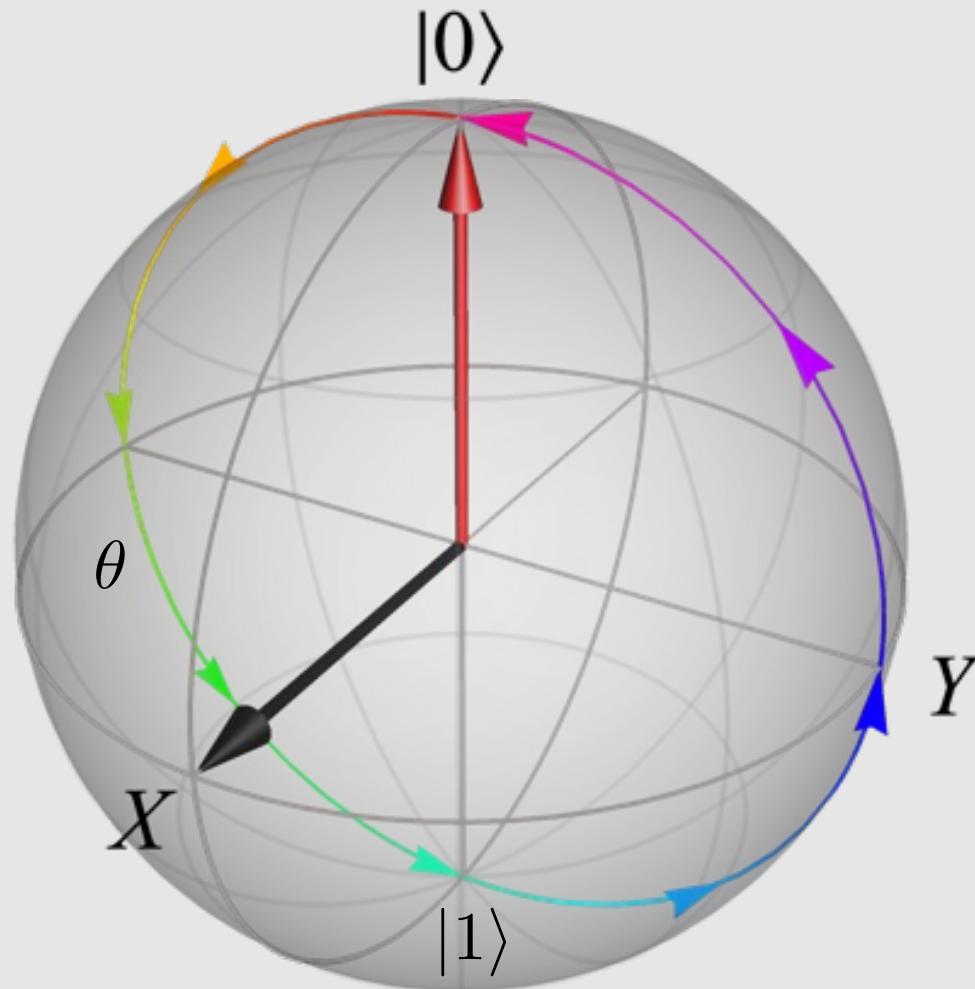


# Visualize: Evolution on the Bloch sphere

$$R_X(\theta) = \exp\left(-\frac{i\theta}{2}X\right)$$

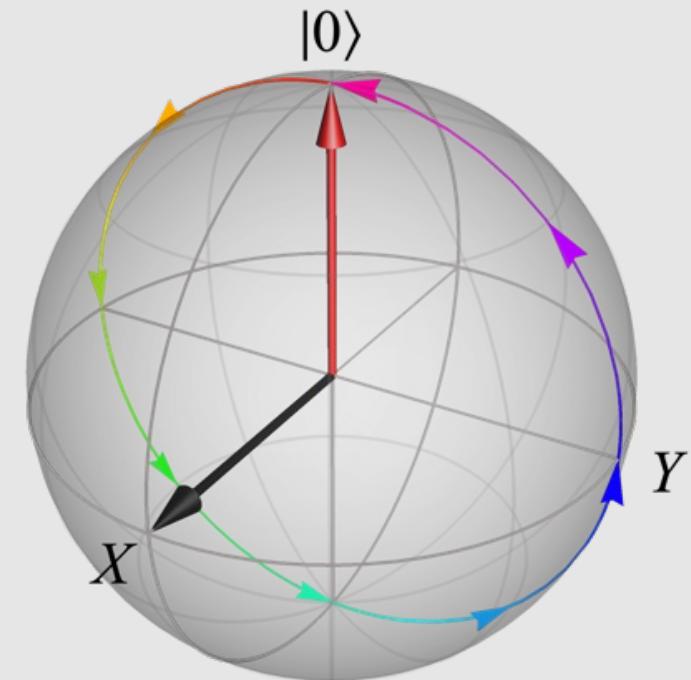
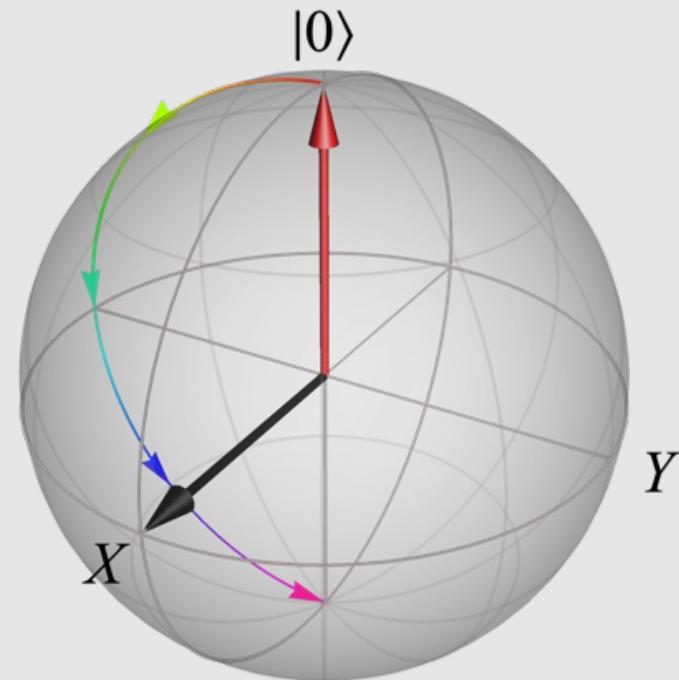
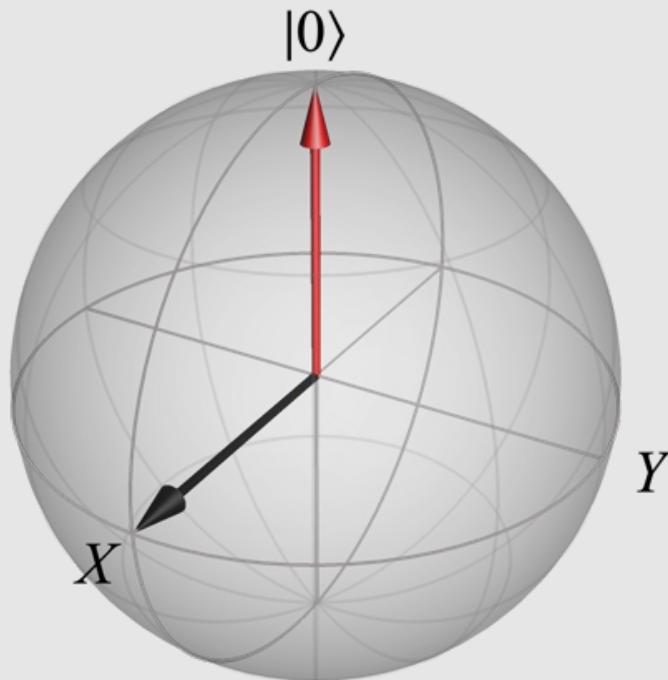
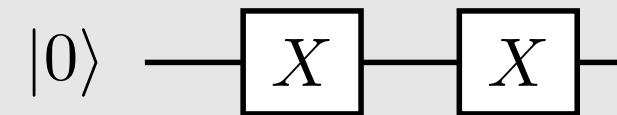
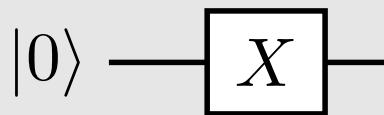
$$X|0\rangle = |1\rangle$$

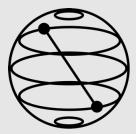
$$X|1\rangle = |0\rangle$$



# Evolution on the Bloch sphere

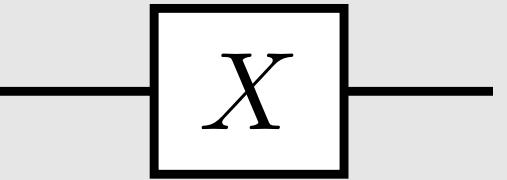
$|0\rangle$



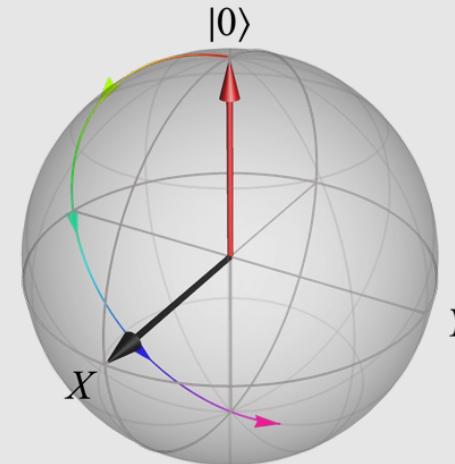
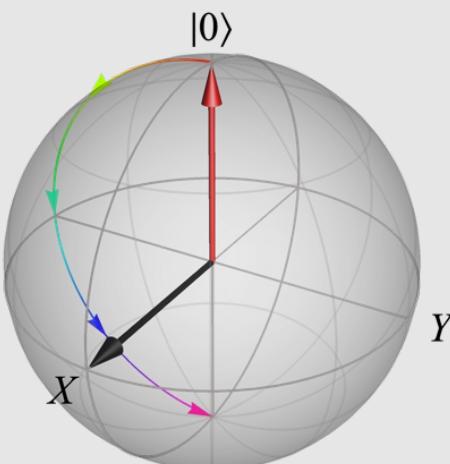
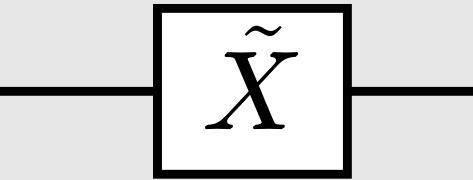


# Miscalibrated gate

Ideal gate

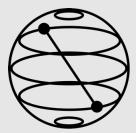


Noisy gate



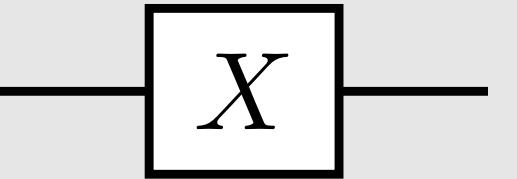
$$X = R_X(\pi)$$

$$\tilde{X} := R_X(\pi + \epsilon)$$

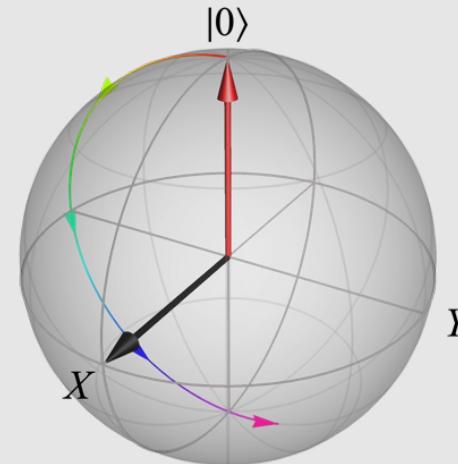
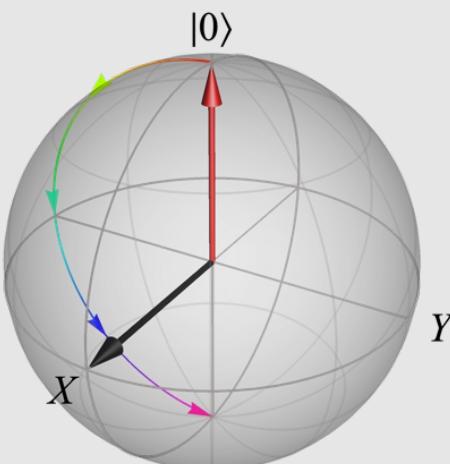
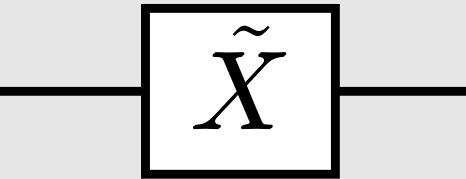


# Miscalibrated gate

Ideal gate



Noisy gate



Shown

$$R_X(\theta + \phi) = R_X(\theta) R_X(\phi)$$

$$X = R_X(\pi)$$

$$\tilde{X} := R_X(\pi + \epsilon)$$

$$R_X(\theta) = \exp\left(-\frac{i\theta}{2}X\right)$$

$$= \cos(\theta/2)I - i \sin(\theta/2)X$$

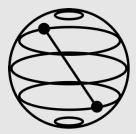
$$= \exp\left(-i\frac{\pi + \epsilon}{2}X\right)$$

$$= \exp\left(-i\frac{\pi}{2}X - i\frac{\epsilon}{2}X\right)$$

$$= \exp\left(-i\frac{\pi}{2}X\right) \exp\left(-i\frac{\pi}{2}X\right)$$

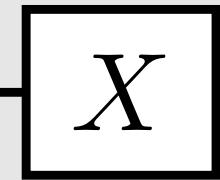
$$= R_X(\epsilon) R_X(\pi)$$

$$= X_\epsilon X$$

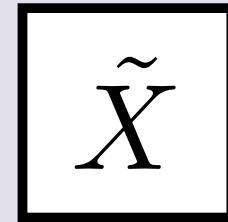
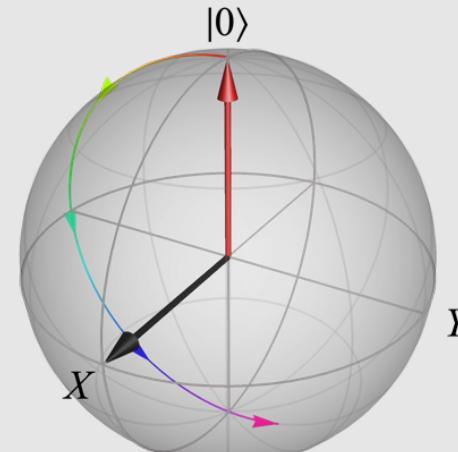
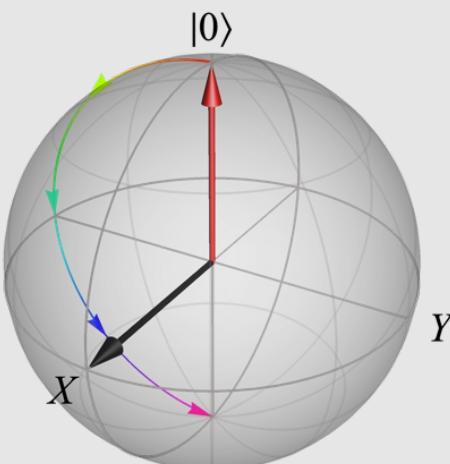
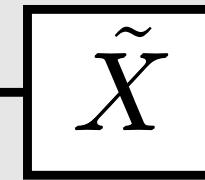


# Noisy gate decomposition

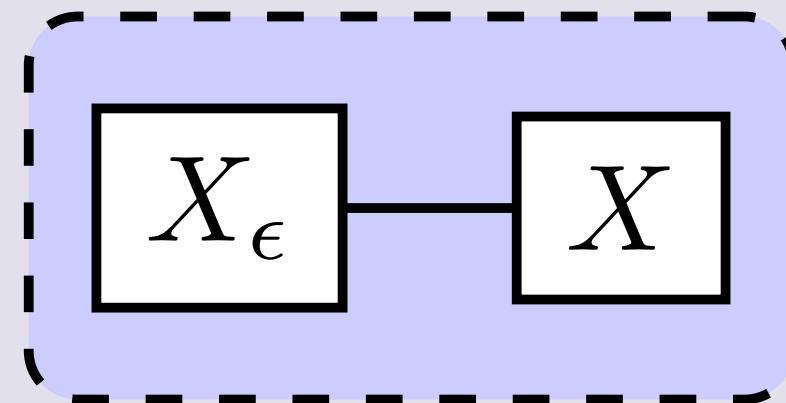
Ideal gate



Noisy gate



||



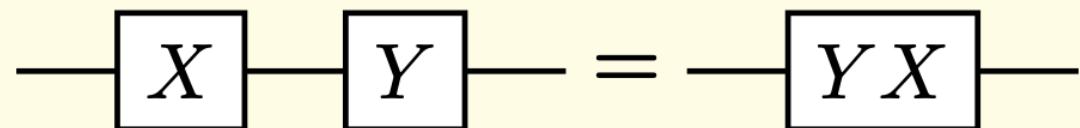


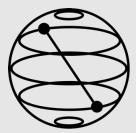
# Careful



## Common pitfall

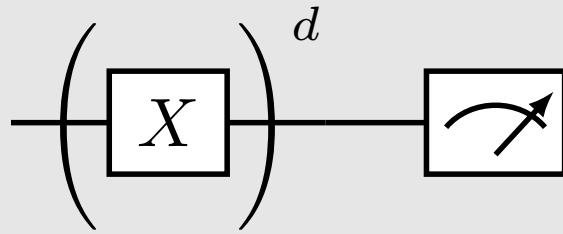
The order in which gates appear in a schematic is the reverse of how they appear in the algebra.





# Using a noisy gate in a quantum circuit

Ideal



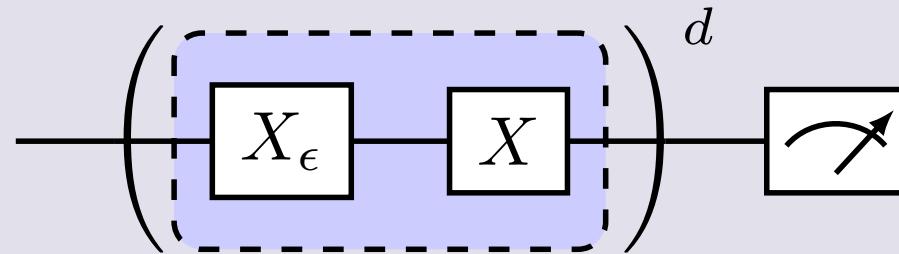
$$X = R_X(\pi)$$

$$U_{\text{total}} = X^d$$

$$= [R_X(\pi)]^d$$

$$= R_X(d\pi)$$

Noisy



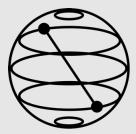
$$\tilde{X} := R_X(\pi + \epsilon) = X_\epsilon X$$

$$\tilde{U}_{\text{total}} = \tilde{X}^d$$

$$= [R_X(\epsilon) R_X(\pi)]^d$$

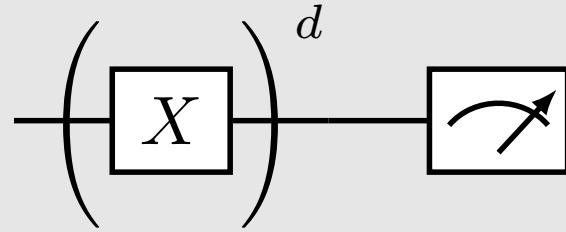
$$= R_X(d\epsilon) R_X(d\pi)$$

$$\tilde{U}_{\text{total}} = R_X(d\epsilon) U_{\text{total}}$$



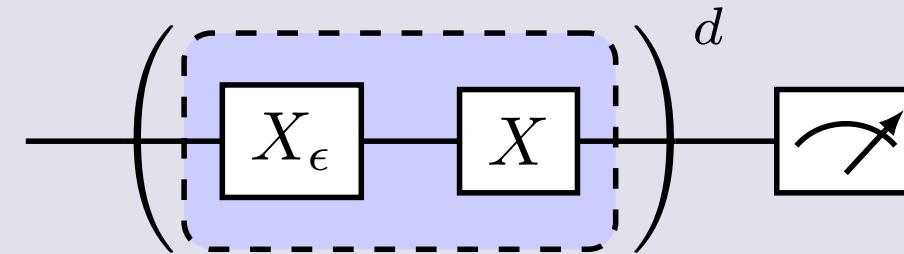
# Using a noisy gate in a quantum circuit: final state

Ideal

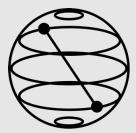


$$U_{\text{total}} = X^d = R_X(d\pi)$$

Noisy

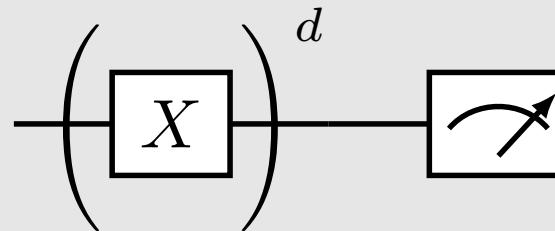


$$\tilde{U}_{\text{total}} = R_X(d\epsilon) U_{\text{total}}$$

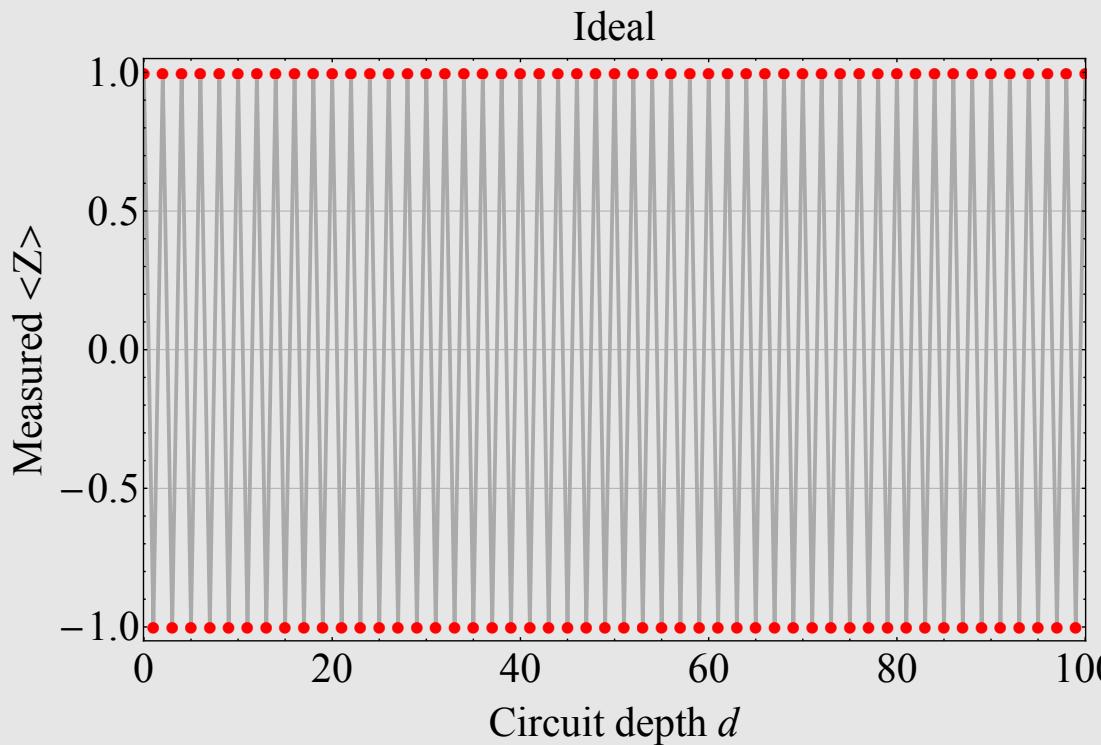


# Ideal vs. noisy observable

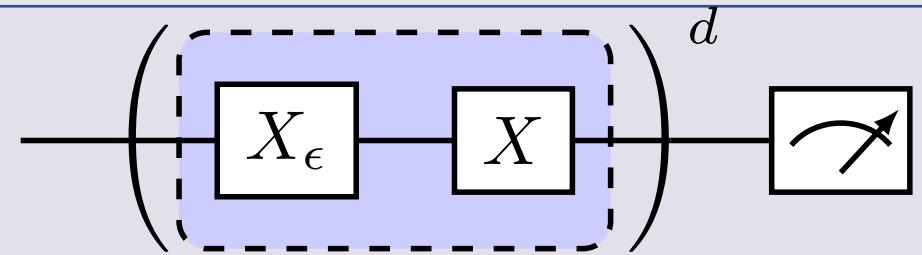
Ideal



$$\langle \psi_f | Z | \psi_f \rangle = \cos(d\pi) = (-1)^d$$

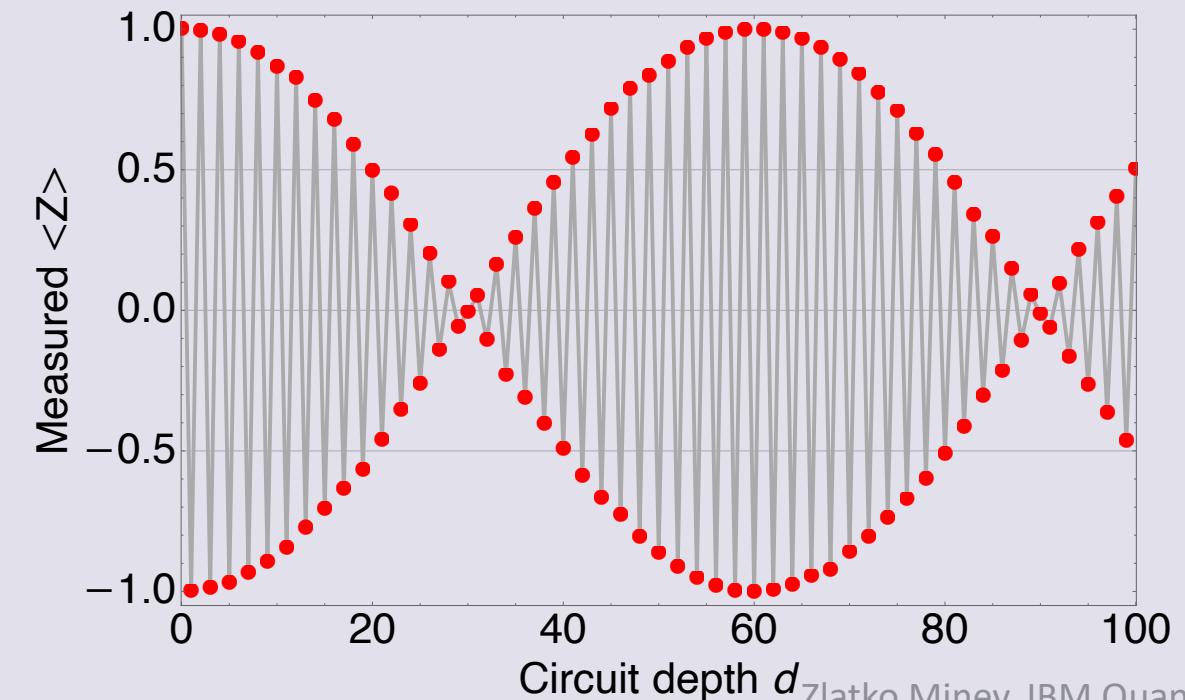


Noisy



$$\langle \tilde{\psi}_f | Z | \tilde{\psi}_f \rangle = \cos(d\pi + d\epsilon)$$

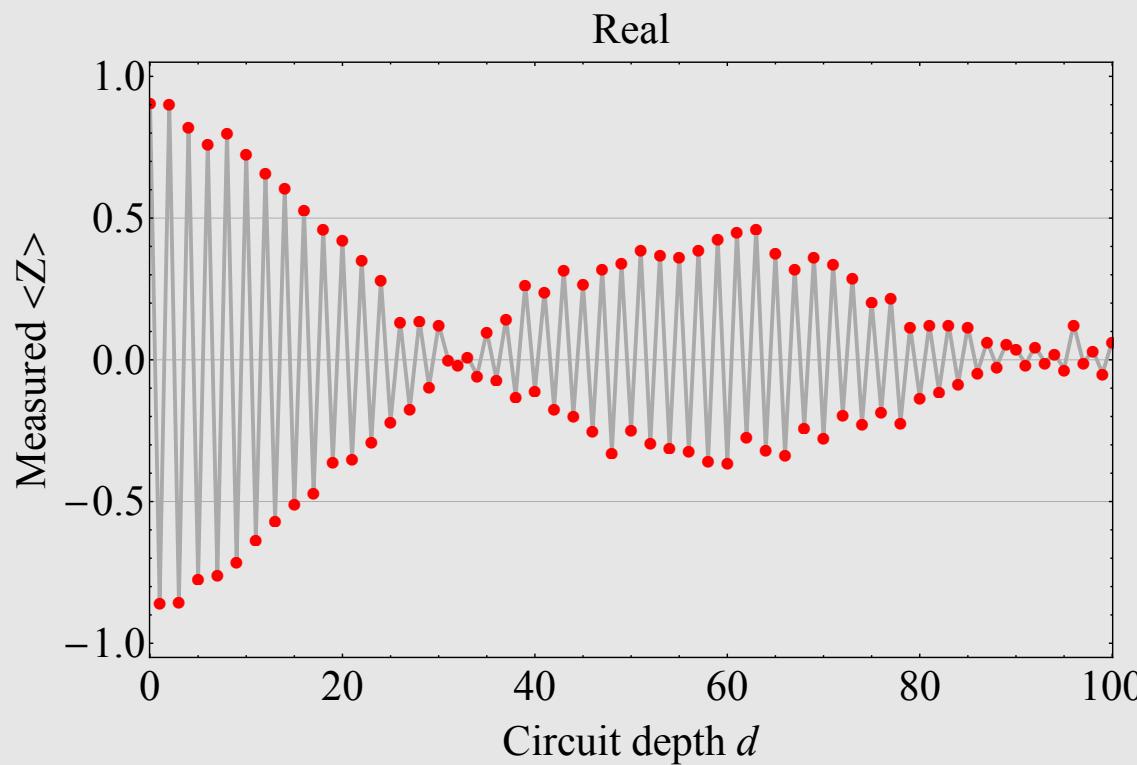
Gate error  $\epsilon = 3^\circ$



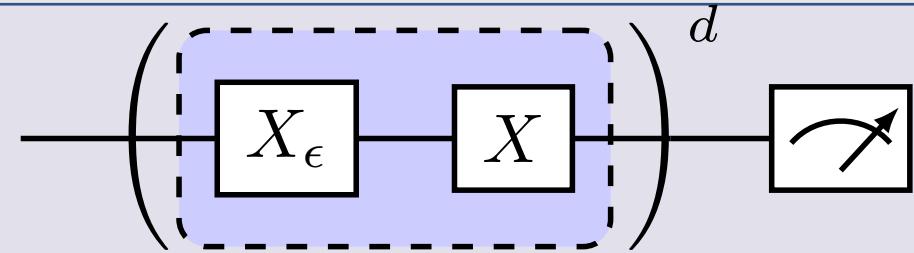


# Compare to full experiment

Full experiment

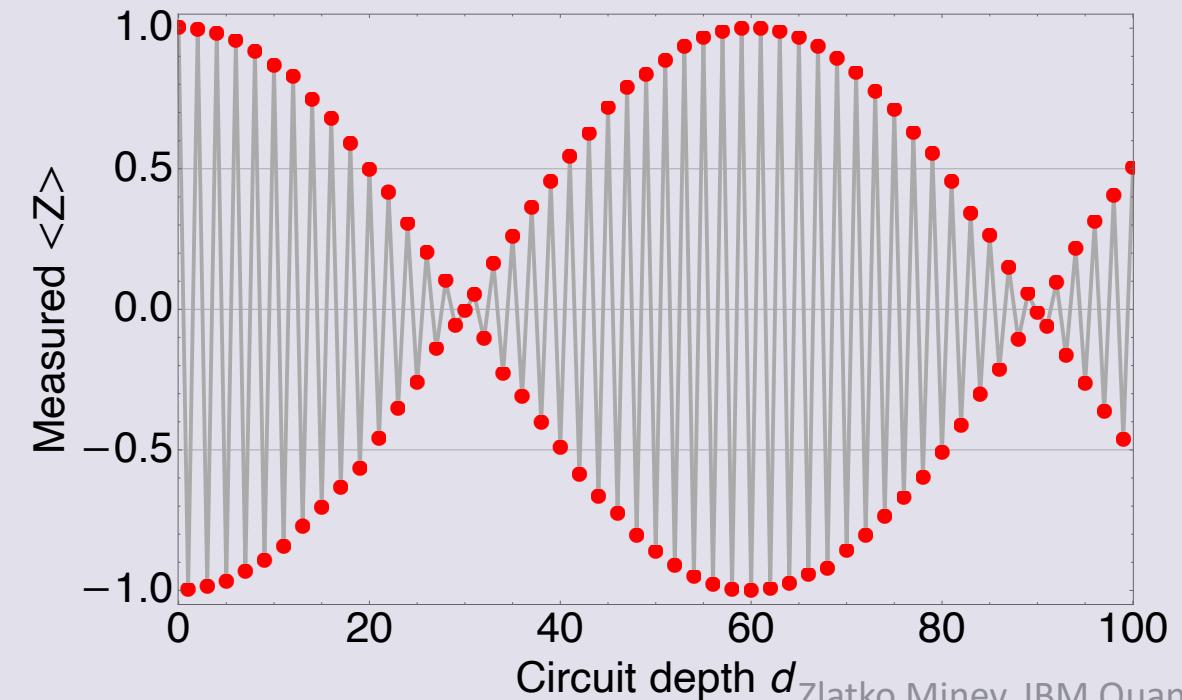


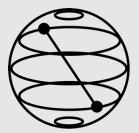
Noisy



$$\langle \tilde{\psi}_f | Z | \tilde{\psi}_f \rangle = \cos(d\pi + d\epsilon)$$

Gate error  $\epsilon=3^\circ$



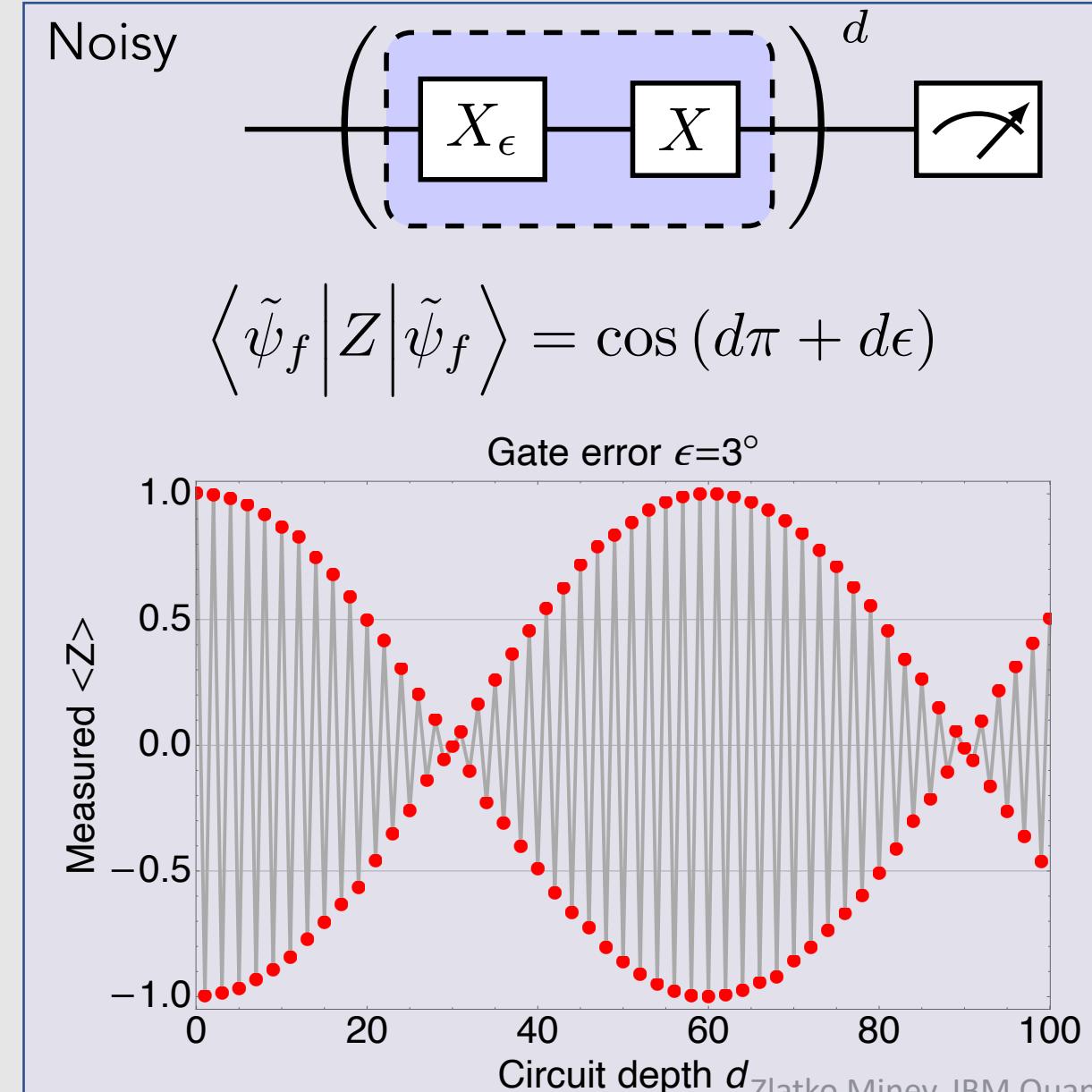


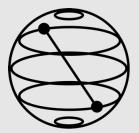
# Coherent error is bad, quadratically so

Coherent errors have a quadratic impact on algorithmic accuracy (worst-case error)

$$\langle \tilde{\psi}_f | Z | \tilde{\psi}_f \rangle - \langle \psi_f | Z | \psi_f \rangle$$

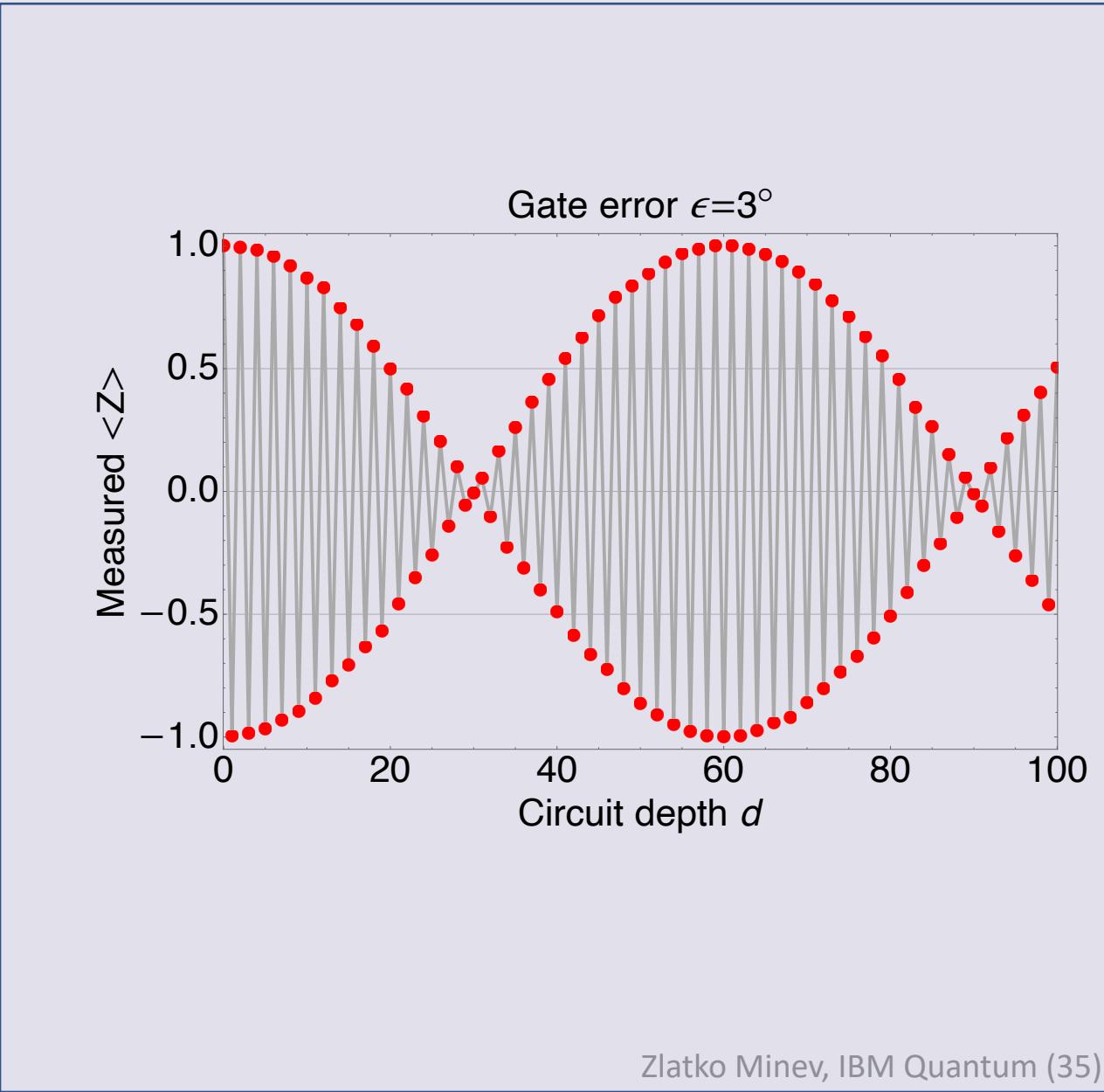
$$\cos(x) = 1 - \frac{1}{2}x^2 + \mathcal{O}(x^3)$$



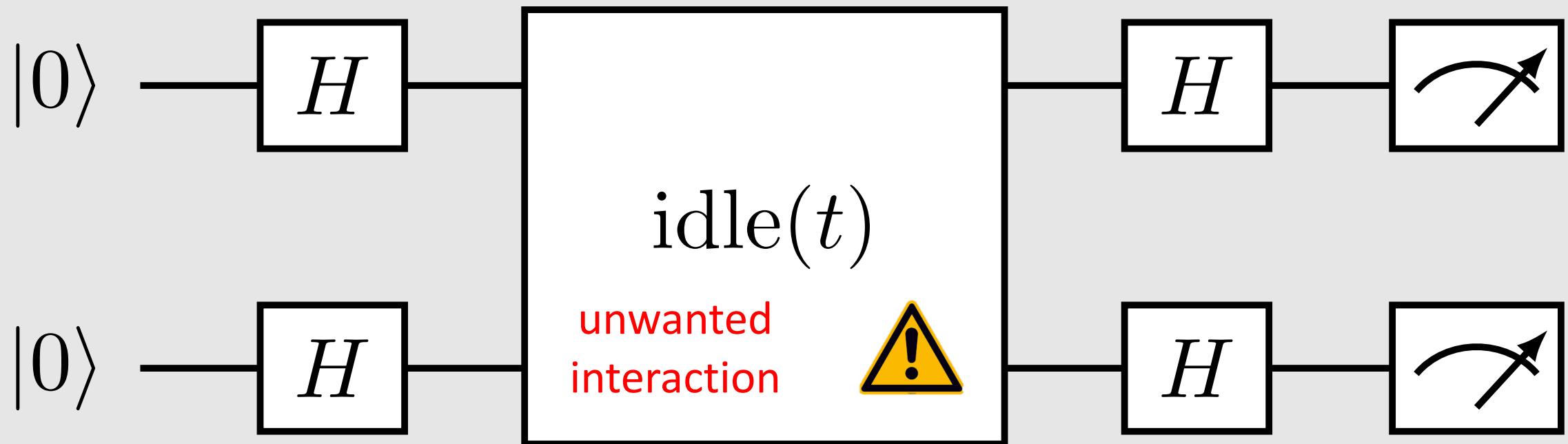


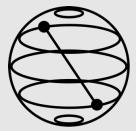
# Coherent errors: brief summary

- are ubiquitous
- can be described by unitary operations
- do not loose quantum information
- data: can create oscillations in the data
- data: do not yield exponential decays
- have a *quadratic* impact on algorithmic accuracy (worst-case error)



## Bonus content: two-qubit coherent ZZ error





# Questions

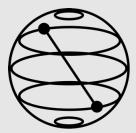
Answer these multiple-choice questions  
in the chat; for example, type “1a 2b.”

1. Coherent noise can be caused by

- a) loss of energy of the qubit
- b) miscalibration, such as over-rotation
- c) wanted coupling to neighboring qubit

2. Coherent noise can be really bad because

- a) it results in loss of information
- b) you cannot undo it
- c) the worst-case error often grows quadratically



# Advanced questions to dive deeper



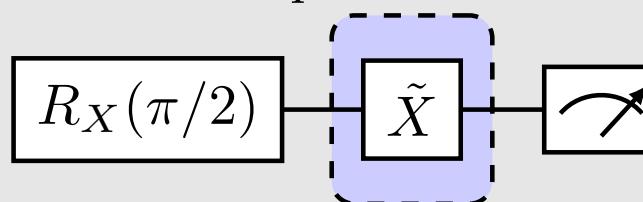
## 1. Amplitude-error amplification sequences.

(A) Calculate the expectation value  $\langle Z(d) \rangle$  as a function of depth  $d$  of the sequence for the following sequences. Assume the noise model  $\tilde{X} := R_X(\pi + \epsilon) = X_\epsilon X$

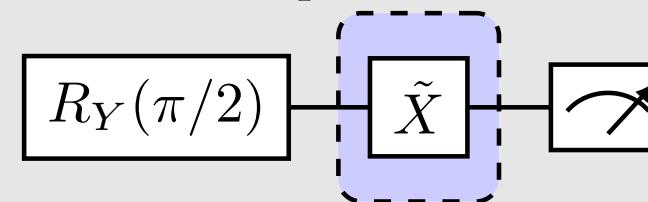
(B) How could you use this result to fine-tune your gates?

(C) Can you come up with an alternative or more clever error-amplification sequence?

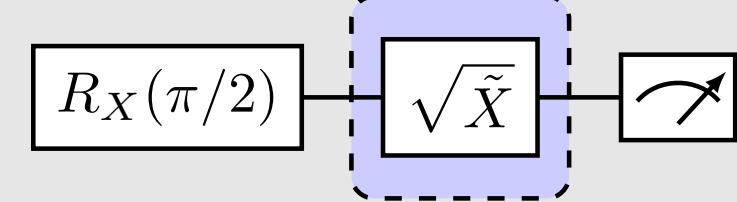
(a) repeat  $d$  times



(b) repeat  $d$  times



(c) repeat  $d$  times



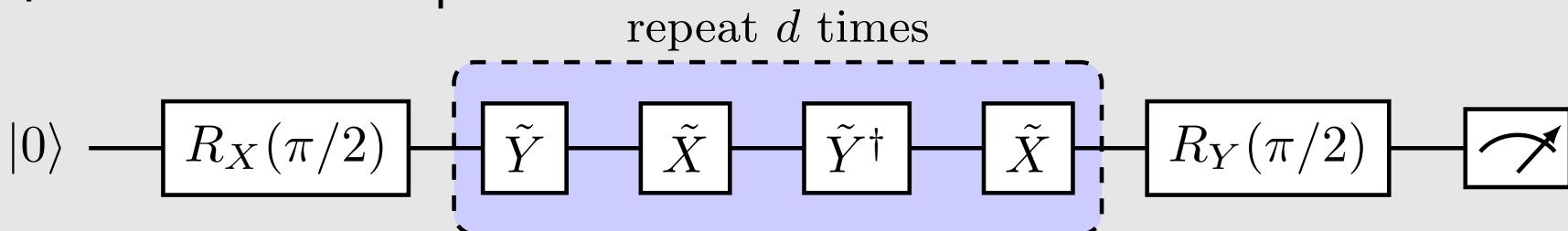
## 2. Phase-error amplification sequences. Use the noise model:

Repeat (A), (B), and (C) for Exercise (1) for the following sequences and assuming a phase error between the  $X$  and  $Y$  gates, rather than an amplitude error.

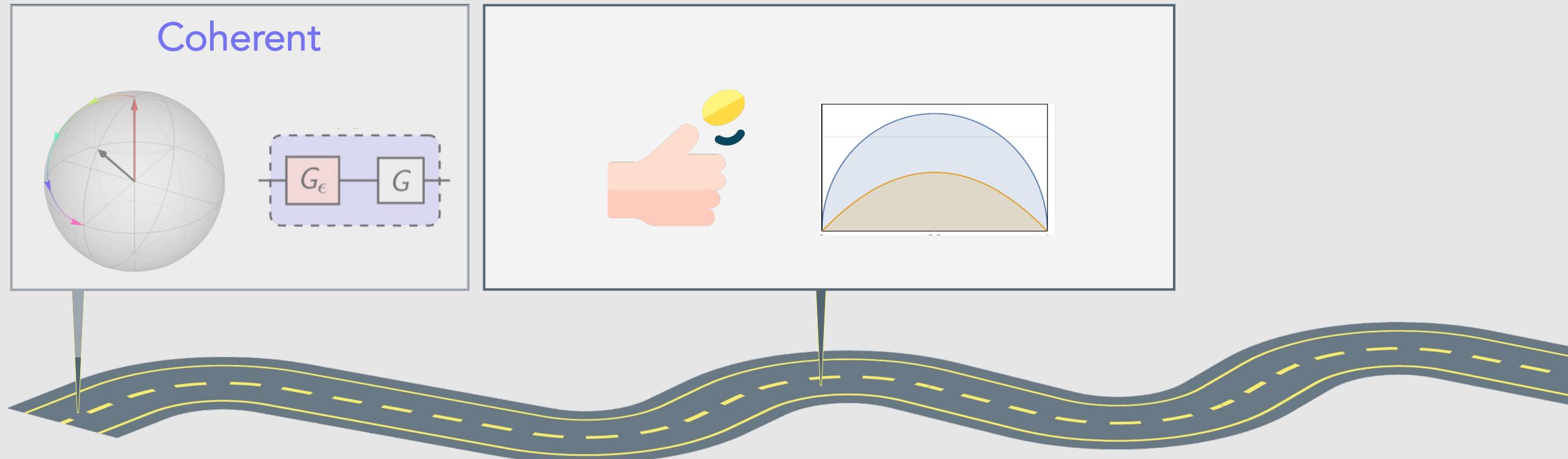
$$\tilde{X} = X ,$$

$$\tilde{Y} = \exp \left[ -i \frac{\pi}{2} (\cos \epsilon Y + \sin \epsilon X) \right] .$$

repeat  $d$  times



# Chapter 3: Measurement theory in a nutshell + projection noise

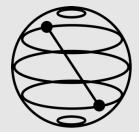


coin toss: flaticon; spam: make it move;  
road based on: freepik

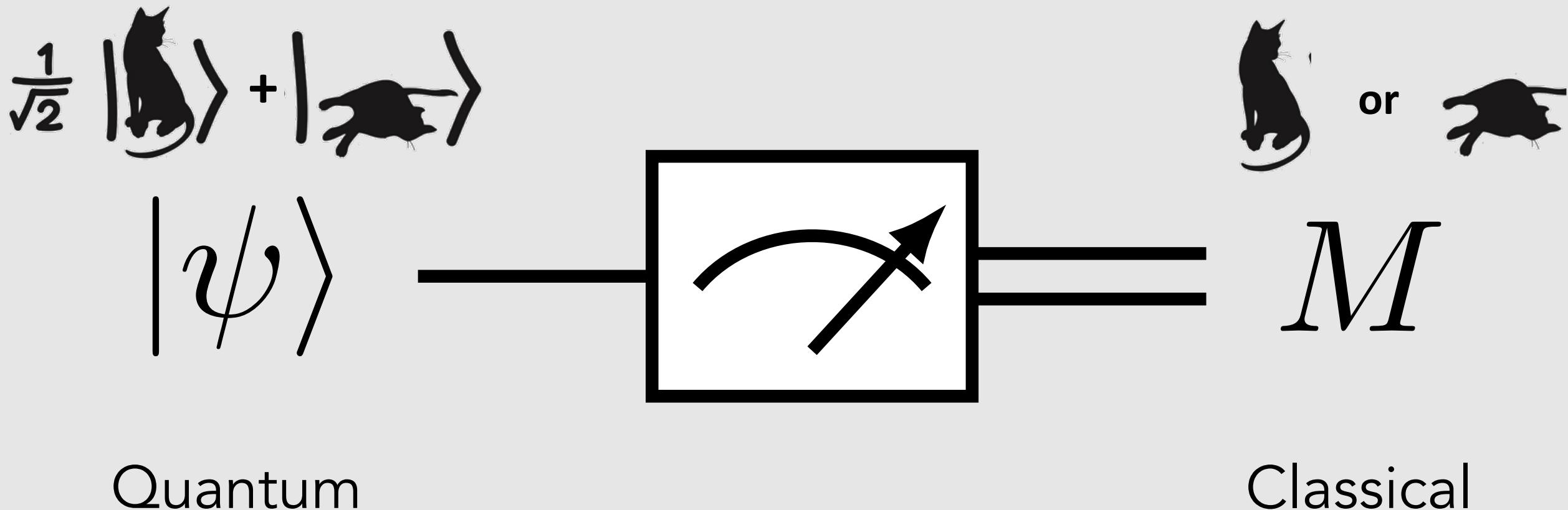
# Quantum Measurement

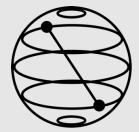


Bohr and Einstein (1925)  
Source: Wikimedia

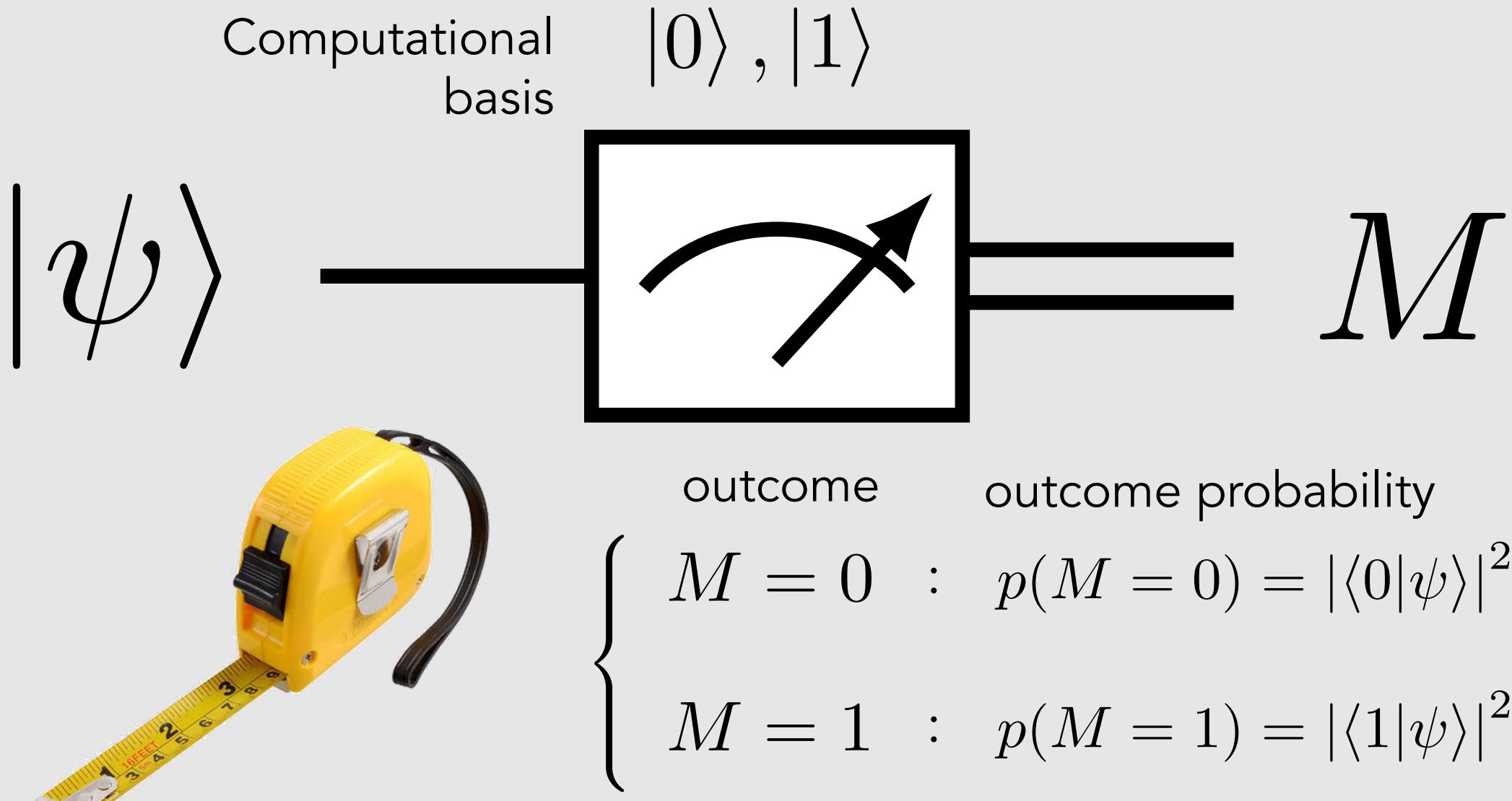


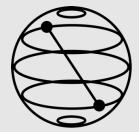
# Measurement apparatus: general



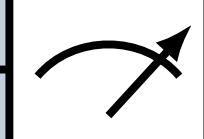


# Measurement basis



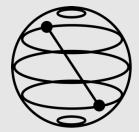


# Resolution of the identity

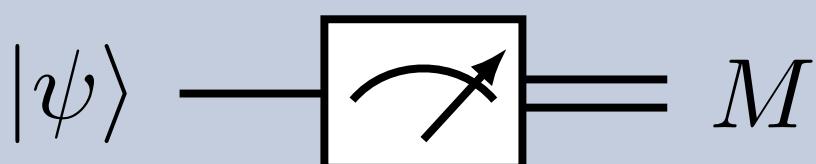
$|\psi\rangle$  —————  =  $M$

outcome      outcome probability

$$\left\{ \begin{array}{l} M = 0 : p(M = 0) = |\langle 0|\psi \rangle|^2 \\ M = 1 : p(M = 1) = |\langle 1|\psi \rangle|^2 \end{array} \right.$$



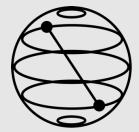
# Resolution of the identity



outcome      outcome probability

$$\begin{cases} M = 0 & : p(M = 0) = |\langle 0|\psi \rangle|^2 \\ M = 1 & : p(M = 1) = |\langle 1|\psi \rangle|^2 \end{cases}$$

$$\begin{aligned} p(M = 0) + p(M = 1) &= |\langle 0|\psi \rangle|^2 + |\langle 1|\psi \rangle|^2 \\ &= \langle \psi|0\rangle \langle 0|\psi \rangle + \langle \psi|1\rangle \langle 1|\psi \rangle \\ &= \langle \psi|(|0\rangle \langle 0| + |1\rangle \langle 1|)|\psi \rangle \\ &= \langle \psi| \hat{I} |\psi \rangle \\ &= ||\psi\rangle|^2 \\ &= 1 \end{aligned}$$



# Projects and overlaps

$$|\psi\rangle \xrightarrow{\text{ } \square \text{ }} M$$

outcome	outcome probability
$M = 0$	$: p(M = 0) =  \langle 0 \psi \rangle ^2$
$M = 1$	$: p(M = 1) =  \langle 1 \psi \rangle ^2$

$$p(M = 0) + p(M = 1)$$

$$= \langle \psi | (|0\rangle\langle 0| + |1\rangle\langle 1|) |\psi \rangle$$

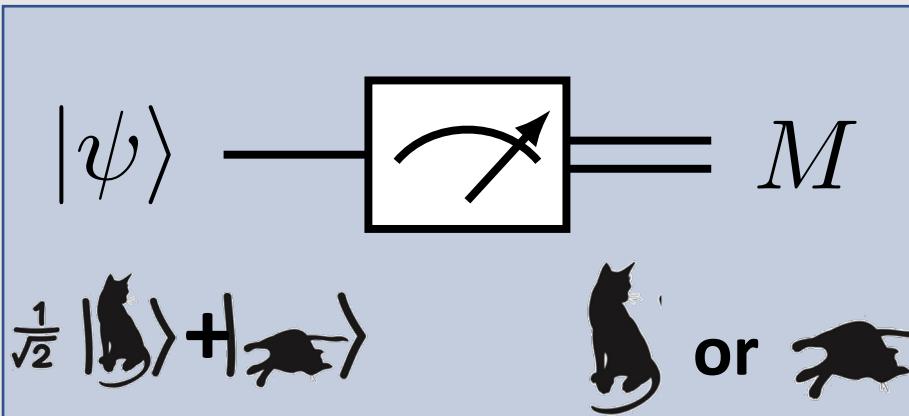
projectors

$$\hat{\Pi}_0 = |0\rangle\langle 0| \quad \hat{\Pi}_1 = |1\rangle\langle 1|$$

inner-product in operator space (overlap between states)

$$p(m) = \langle \hat{\Pi}_m, |\psi \rangle \rangle = \text{Tr} (\hat{\Pi}_m |\psi \rangle \langle \psi |)$$

# Measurement theory 101 summary: Scaffolding in terms of measurement operators



**Measurement aspect**

Measurement outcome  $M$

**Example**

$$M = 0$$

Set of measurement outcomes  $\Sigma$

$$\Sigma = \{0, 1\}$$

Measurement operator  $\Pi_M$

$$\Pi_M = |0\rangle\langle 0|$$

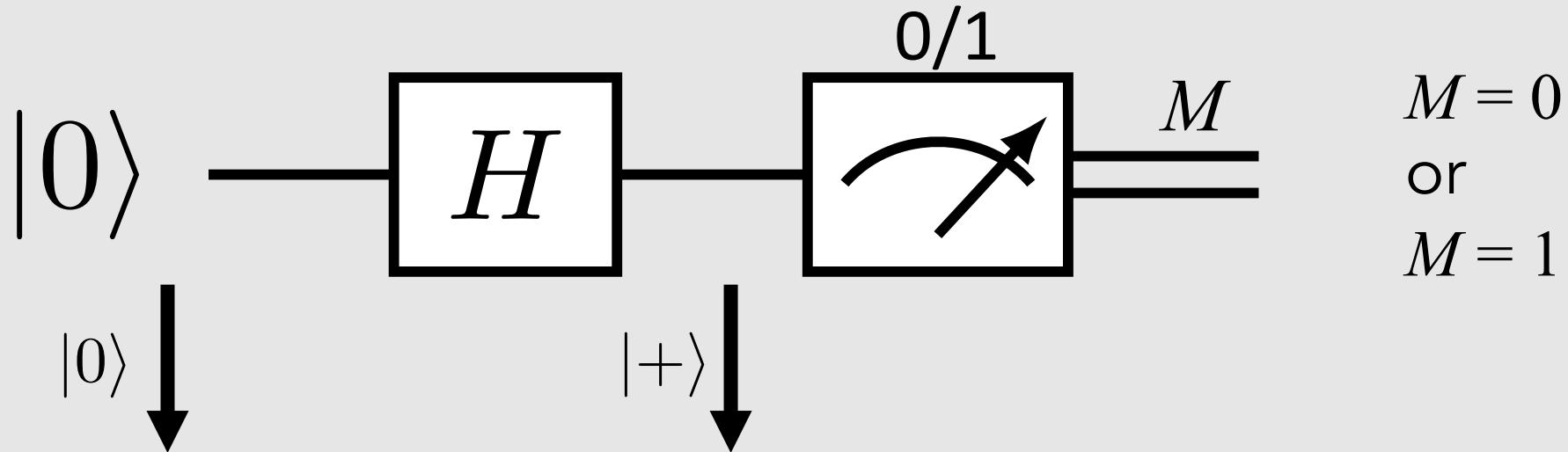
Probability of measuring  $M = m$

$$\begin{aligned} p(M=m) &= \langle \Pi_m, \psi \rangle \\ &= |\langle \Pi_m | \psi \rangle|^2 \end{aligned}$$

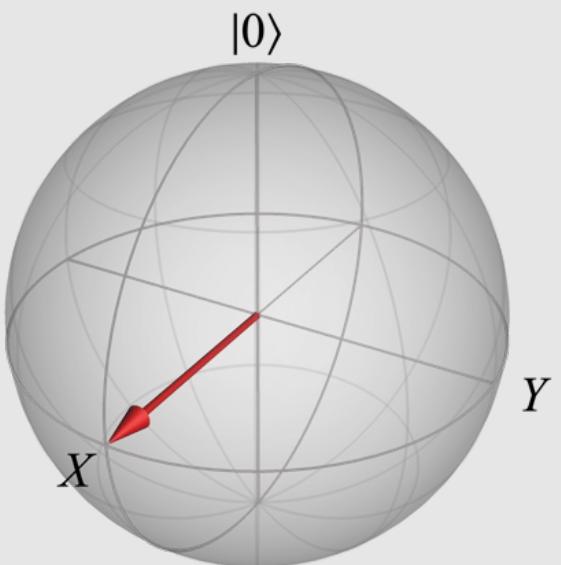
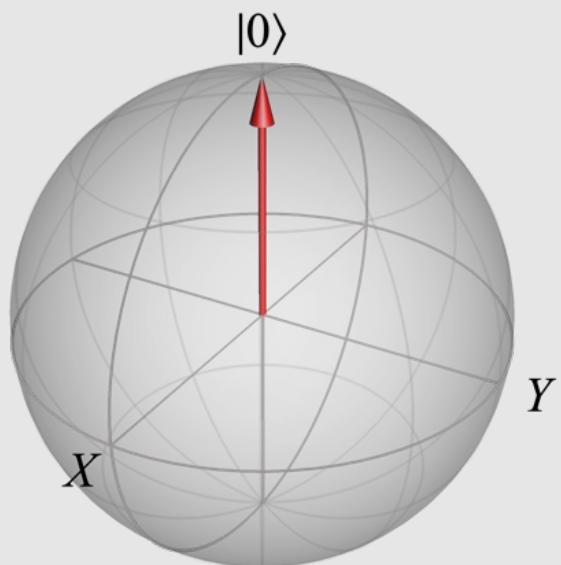




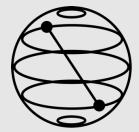
# Putting it to use: example ideal circuit



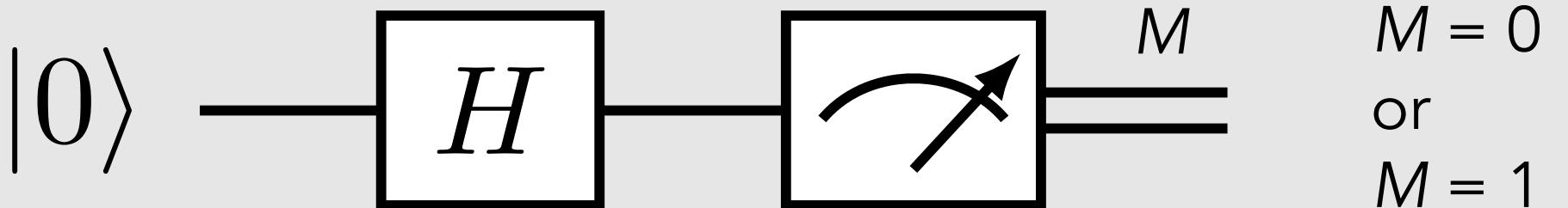
Quantum state



Classical random variable



# Example ideal circuit



more general case

Prob to find  $M=1$  :  $P(M=1) = p$   $p \in [0, 1]$

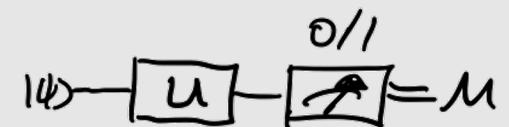
every possible qubit case

Bernoulli Distribution

$P(M) \sim B(p)$

$\uparrow$  classical rand variable

probability



# Mean

## Statistics

$$\mathbb{E}[M] = \sum_m m P(m) = 0 P(m=0) + 1 P(m=1)$$

$\approx P(m=1)$

= p

$$= \sum_m m \langle \hat{\Pi}_m \rangle$$

$$= \sum_m m \langle |m\rangle \langle m| \rangle$$

$$= \langle \sum_m m |m\rangle \langle m| \rangle$$

=  $\langle \hat{M} \rangle$

$$V[M] = \mathbb{E}[M^2] - \underbrace{\mathbb{E}[M]^2}_{P^2} \approx \langle \hat{M}^2 \rangle - \langle \hat{M} \rangle^2$$

$$\begin{aligned}\mathbb{E}[M^2] &= \sum_m m^2 P(m) \\ &= \sum_m m^2 \langle |m\rangle \langle m| \rangle \\ &= \left\langle \sum_m m^2 |m\rangle \langle m| \right\rangle \\ &= \langle \hat{M}^2 \rangle \\ &\approx \cancel{0^2 + p} + l^2 p \\ &\approx p\end{aligned}$$

$$\begin{aligned}V[M] &\approx p - p^2 \\ &\approx p(1-p) \\ &\approx \sigma_M^2\end{aligned}$$

$= 0$  if  $p=0$   
 $= 1$  if  $p=1$

# Different observables

Different observables

For example  $\hat{M} = |0\rangle\langle 0| + |1\rangle\langle 1| = \frac{1}{2}(\hat{I} + \hat{Z}) = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$

$$\hat{M} = (+)|0\rangle\langle 0| - (-)|1\rangle\langle 1| = \hat{Z} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

For  $\hat{M} = \hat{Z}$   $\Sigma = \{-1, +1\}$  Pauli Observables

$$\hat{M} = \hat{X} \quad \Sigma = \{+, -\}$$

$$\hat{\Pi}_+ = |+\rangle\langle +|$$

$$\hat{\Pi}_- = |->\langle -|$$

$$\hat{X} = \bigcup_{m \in \Sigma} m |m\rangle\langle m|$$

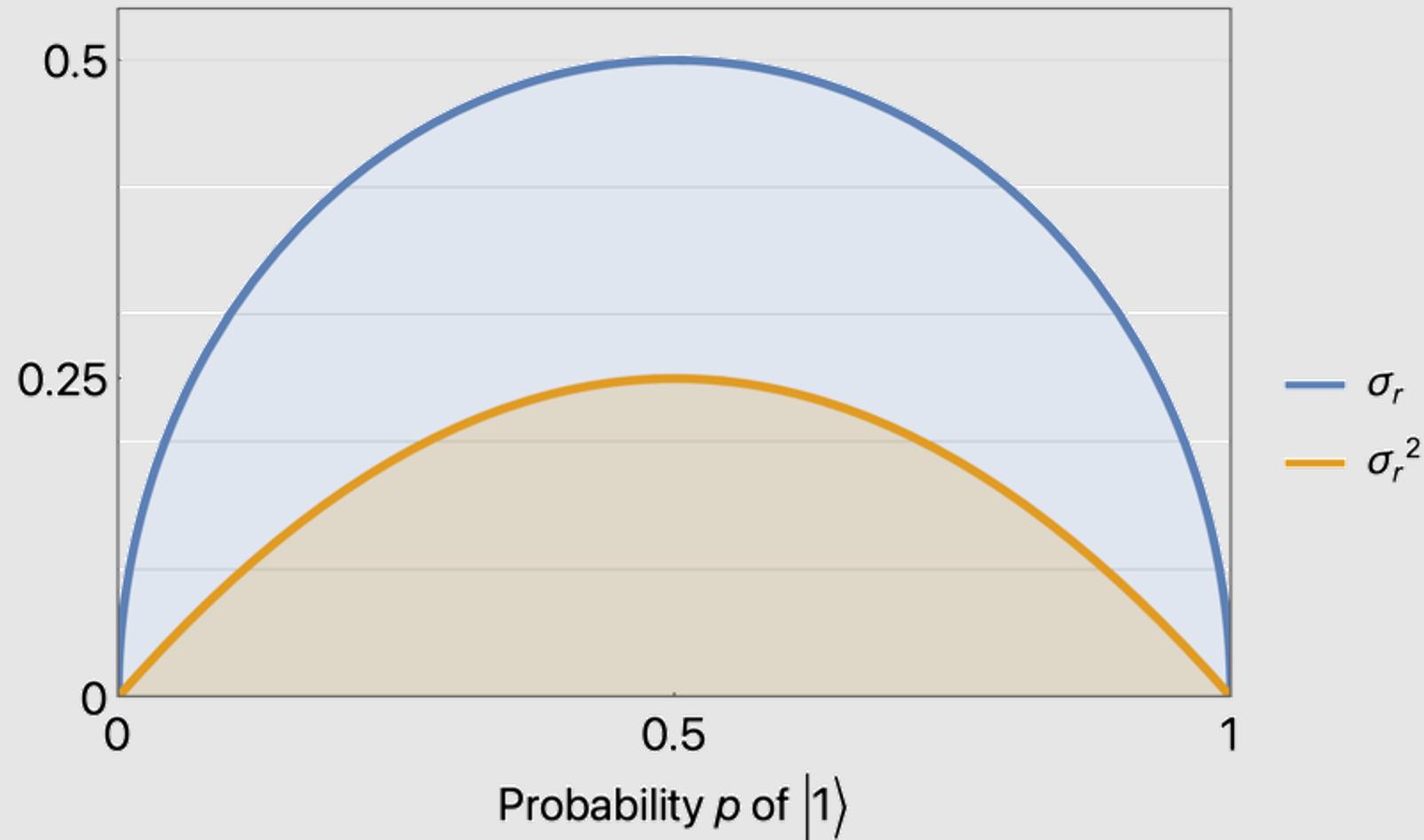
$$\approx |+\rangle\langle +| - |-\rangle\langle -|$$

$$\hat{X}|+\rangle = +|+\rangle$$
$$\hat{X}|-\rangle = -|-\rangle$$

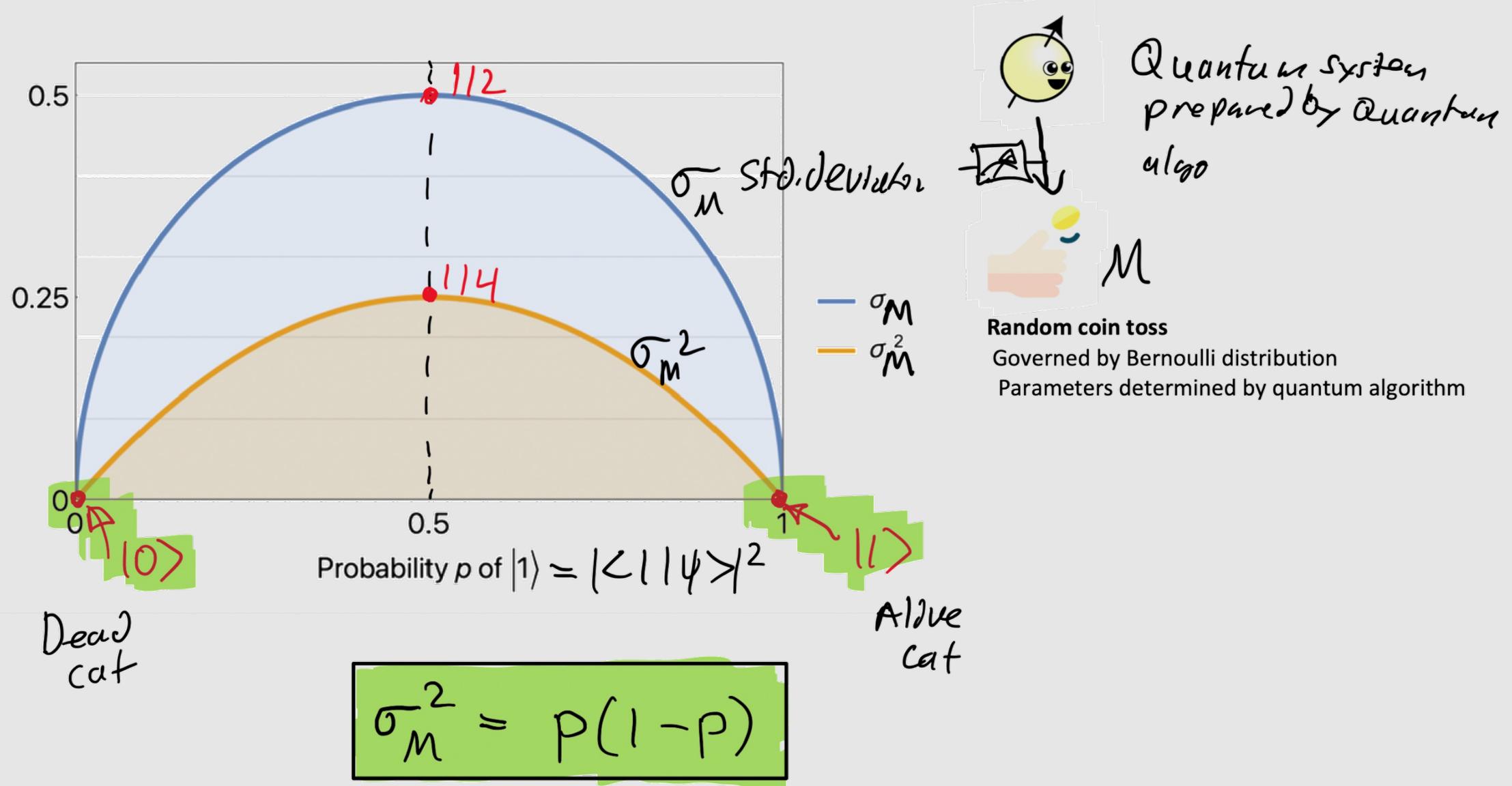
$$|+\rangle = \frac{1}{\sqrt{2}}(|+\rangle)$$

$$|-\rangle = \frac{1}{\sqrt{2}}(|-\rangle)$$

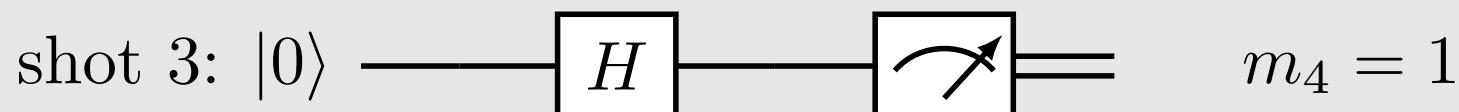
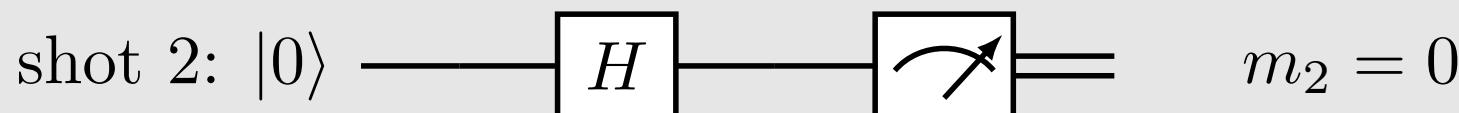
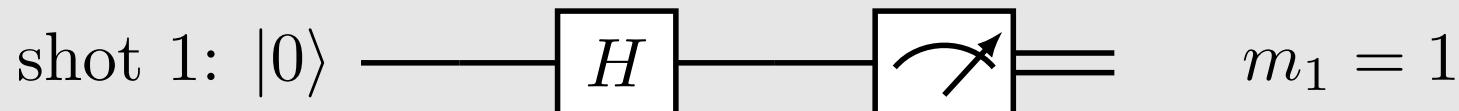
# Variance of the random classical variable vs. probability to obtain 1



# Variance of the random classical variable vs. probability to obtain 1

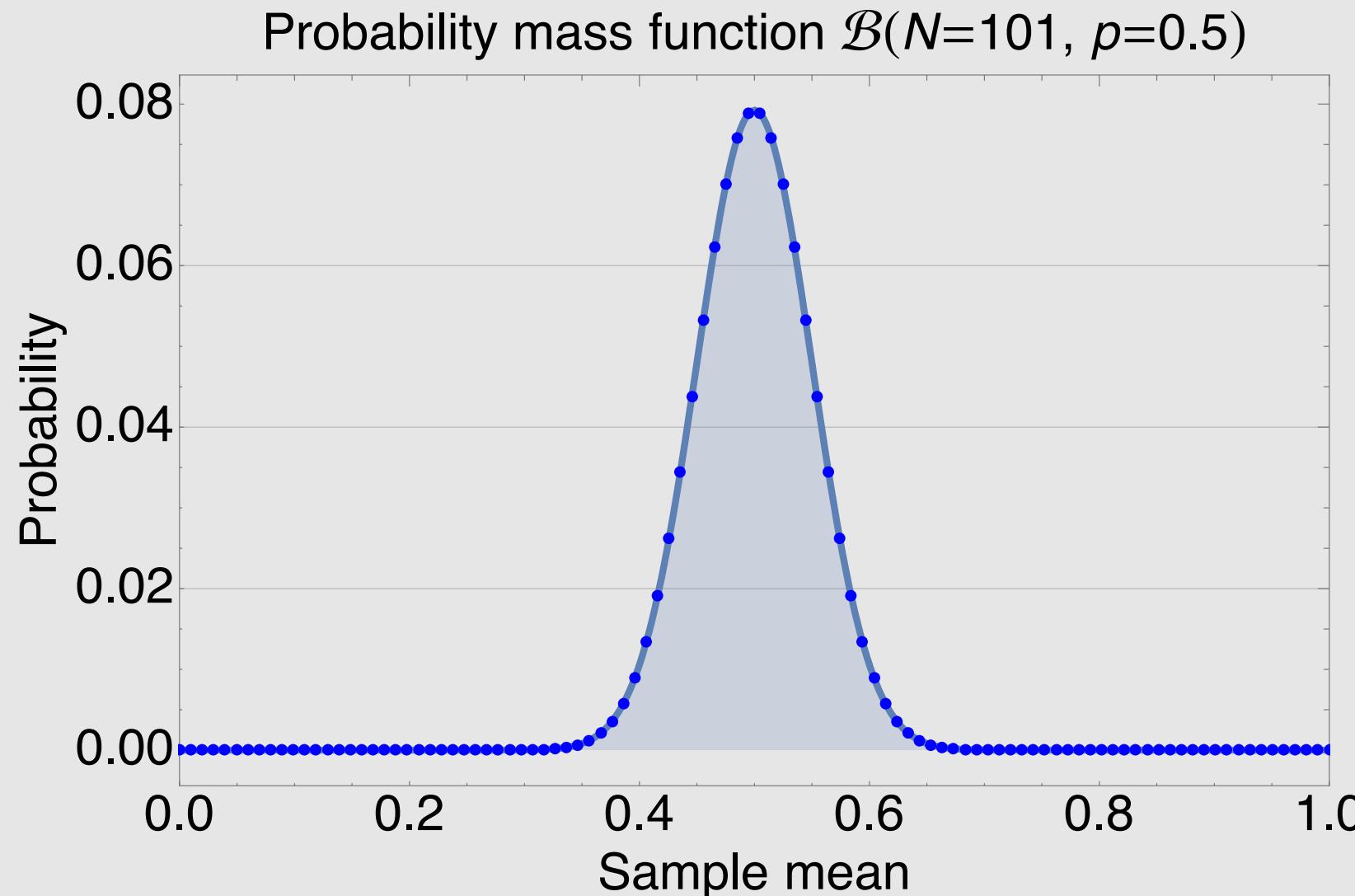


# Shots, shots, shots

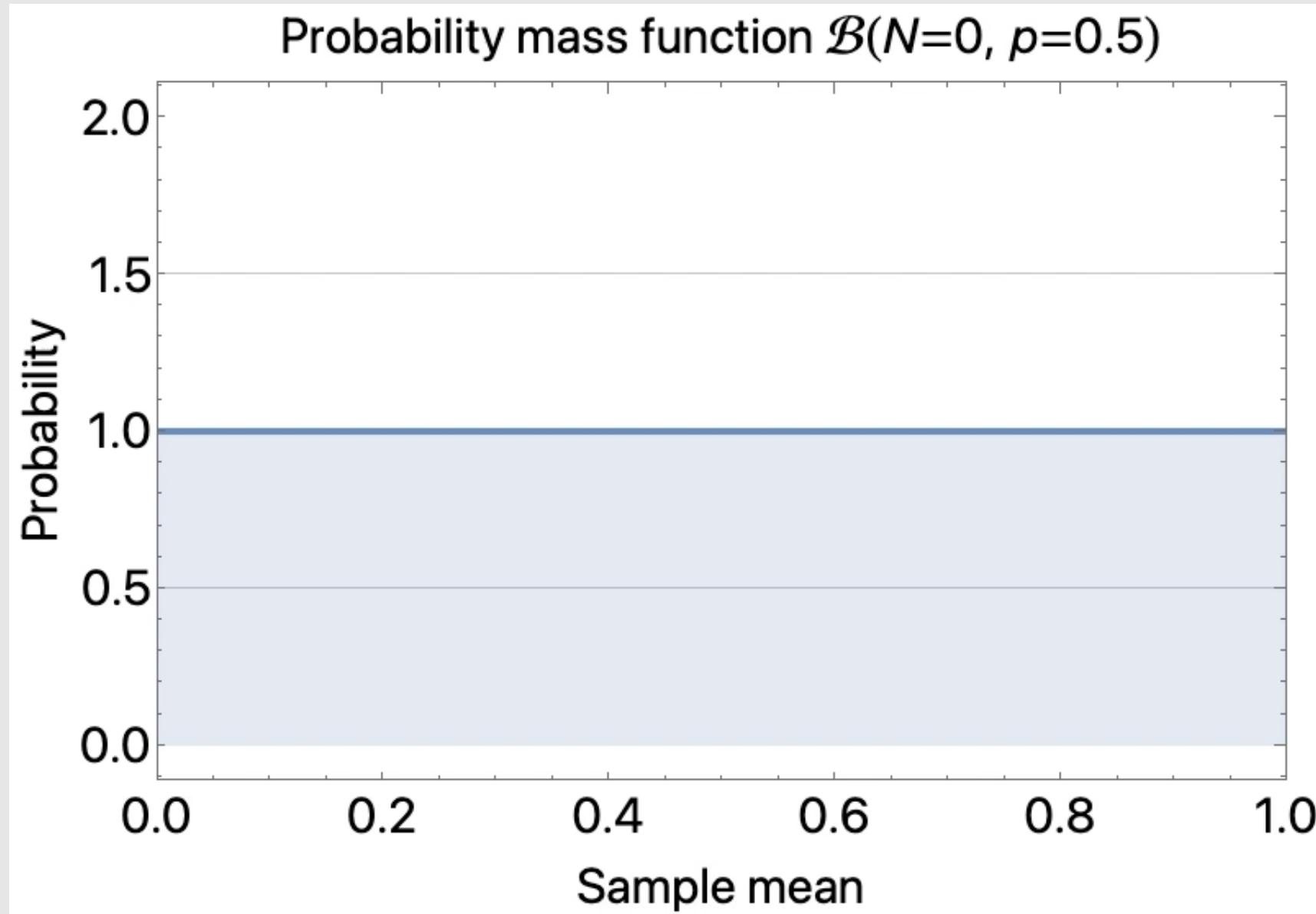


:

# Sampled output distribution



# Animation of convergence of shots expectation value and mean



# Concentration inequalities and tail bounds

Making a list,  
checking it twice,  
going to see  
which inequality  
is nice!

Markov? Hoeffding?  
Jensen? Chebyshev?  
Chernoff?

## 1. Probability (Technical note 11.9 v0.6)

### 1A. Concentration inequalities and tail bounds

Unless otherwise specified, all variables are real  $\mathbb{R}$ . Inequalities come as one-sided  $\Pr(\dots \leq \dots)$  and two-sided  $\Pr(|\dots| \leq \dots)$ . Notation:  $X$  is a random variable,  $\mu := \mathbb{E}[X]$ ,  $\sigma^2 := \text{Var}[X]$ ,  $S_n := X_1 + \dots + X_n$ .

Inequality	Conditions	Common form	Notes / Alternate form
<b>Single random variable</b>			
<b>Markov</b> <sup>1</sup>	Non-negative $X \geq 0$	$\Pr[X \geq a] \leq \frac{\mathbb{E}[X]}{a}$	$\forall a > 0$ $\Pr[X \geq k\mathbb{E}[X]] \leq \frac{1}{k}$ $k > 1$ [3, Sec. 5.1][6, Thm 1.13]
extension	+ non-negative, strictly increasing func $\Phi$ $X \geq 0$ $\Phi(X) \geq \Phi(a)$ increasing	$\Pr[X \geq a] = \Pr[\Phi(X) \geq \Phi(a)] \leq \frac{\mathbb{E}(\Phi(X))}{\Phi(a)}$	$\forall a > 0$ [Wiki]
<b>Reverse Markov</b>	upper-bounded by $U$ $\max X = U$ (can be positive)	$\Pr[X \leq a] \leq \frac{U - \mathbb{E}[X]}{U - a}$	$\forall a > 0$ [1, Sec. 3.1]
<b>Chebyshev</b> <sup>2</sup>	Finite mean and variance $\mathbb{E}[X]$ , $\text{Var}[X]$ finite	$\Pr[ X - \mathbb{E}[X]  \geq a] \leq \frac{\sigma^2}{a^2}$	$\Pr[ X - \mathbb{E}[X]  \geq a \cdot \sigma] \leq \frac{1}{a^2}$ [1, Sec. 3.2] $\forall a > 0$ , $\sigma^2 = \text{Var}[X]$ [3, Sec. 5.1][2, Thm 18.11]
<b>Cantelli</b>	Improved Chebyshev (same; but one-sided)	$\Pr[X - \mathbb{E}[X] \geq a] \leq \frac{\sigma^2}{\sigma^2 + a^2}$	$\forall a > 0$ , $\sigma^2 = \text{Var}[X]$ [Wiki]
<b>Chernoff</b> <sup>3</sup>	Generic	$\Pr[X \geq a] = \Pr[e^{tX} \geq e^{ta}]$	$\forall t > 0$ , $a \in \mathbb{R}$ [1, Sec. 3.3]
<b>Jensen</b>	$f: \mathbb{R} \rightarrow \mathbb{R}$ ; $f$ convex	$f(\mathbb{E}[X]) \leq \mathbb{E}[f(X)]$	[3, Prob. 5.3][6, Thm 1.14]
<b>Hoeffding's lemma</b>	$\mathbb{E}[X] = \mu$ $a \leq X \leq b$	$\mathbb{E}[e^{\lambda X}] \leq e^{\lambda \mu} e^{\frac{\lambda^2(b-a)^2}{8}}$	$\lambda \in \mathbb{R}$ [1, Sec. 3.4]
<b>Sum of random variables</b>			
<b>Chernoff-Hoeffding (one-sided)</b>	$n$ independent random vars $X_1, \dots, X_n$ indep $S_n = X_1 + \dots + X_n$ $X_i \in [a_i, b_i] \quad \forall i$	$\Pr[S_n - \mathbb{E}[S_n] \geq t] \leq \exp\left(\frac{-2t^2 n^2}{\sum_{i=1}^n (b_i - a_i)^2}\right)$	[1, Sec. 3.5]
(two-sided) <sup>4</sup>	(same as above)	$\Pr[ S_n - \mathbb{E}[S_n]  > t] \leq 2 \exp\left(\frac{-2t^2}{\sum_{i=1}^n (b_i - a_i)^2}\right)$	$\forall t \in (0, \frac{1}{2})$ [5, Thm.1.1]
(two-sided iid)	same plus iid, range, mean $\mu$ for each $X_1, \dots, X_n \in [0, 1]$ $\mathbb{E}[X_i] = \mu$ iid	$\Pr\left[\left \frac{S_n}{n} - \mu\right  \geq \epsilon\right] \leq 2 \exp(-2n\epsilon^2)$	$\forall \epsilon > 0$ [6, Thm 1.16]
<b>Thm 1.3</b>	$n$ independent random vars $X_1, \dots, X_n$ indep $S_n = X_1 + \dots + X_n$	$\Pr[S_n - \mathbb{E}[S_n] > \epsilon] \leq 2 \exp\left(\frac{-\epsilon^2}{4 \sum_{i=1}^n \text{Var}[X_i]}\right)$	$\epsilon \in (0, 2 \text{Var}[S_n] / (\max_i  X_i - \mathbb{E}[X_i] ))$ [5, Thm. 1.3]
<b>Azuma</b>			
<b>Weak law of large numbers</b>	$n$ independent iid random vars $X_1, \dots, X_n$ indep $\mathbb{E}[X_i] = \mu$ iid	$\lim_{n \rightarrow \infty} \Pr\left[\left \frac{1}{n} S_n - \mu\right  \geq \epsilon\right] = 0$	$\forall \epsilon > 0$ [3, Sec. 5.2][6, Thm 1.15]
<b>Strong law of large numbers</b>	(same)	$\Pr\left[\lim_{n \rightarrow \infty} \frac{1}{n} S_n = \mu\right] = 1$	[3, Sec. 5.5]
<b>Advanced</b>			
<b>Bennett</b>	$n$ independent zero-mean $X_1, \dots, X_n$ indep $\mathbb{E}[X_i] = 0$ iid	$\Pr[S_n > \epsilon] \leq \exp\left(-n\sigma^2 h\left(\frac{\epsilon}{n\sigma^2}\right)\right)$	$\sigma^2 := \frac{1}{n} \sum_{i=1}^n \text{Var}[X_i]$ , $\forall \epsilon > 0$ , $h(a) := (1+a) \log(1+a) - a$ for $a \geq 0$ [1, 4.1]
<b>Bernstein</b>	(same)	$\Pr[S_n > \epsilon] \leq \exp\left(\frac{-n\epsilon^2}{2(\sigma^2 + \epsilon/3)}\right)$	(same) [1, 4.2]
<b>Efron-Stein</b>	scalar func of vars $f: \chi^n \rightarrow \mathbb{R}$ $X_1, \dots, X_n$ indep w/ values in set $\chi$	$\text{Var}[Z] \leq \sum_{i=1}^n \mathbb{E}\left[(Z - \mathbb{E}_i[Z])^2\right]$	$Z := g(X_1, \dots, X_n)$ $\mathbb{E}_i[Z] := \mathbb{E}[Z   X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_n]$ [1, 4.3]
<b>McDiarmid's</b>	scalar func of vars $f: \chi^n \rightarrow \mathbb{R}$ $X_1, \dots, X_n$ indep w/ values in set $\chi$	$\Pr[f(X_1, \dots, X_n) - \mathbb{E}[f(X_1, \dots, X_n)] \geq \epsilon] \leq \exp\left(\frac{-2\epsilon^2}{\sum_{i=1}^n c_i^2}\right)$	condition: $c$ -bounded difference property $\forall \epsilon > 0$ $ f(X_1, \dots, X_i, \dots, X_n) - f(X_1, \dots, X'_i, \dots, X_n)  \leq c_i$ [1, 4.4]

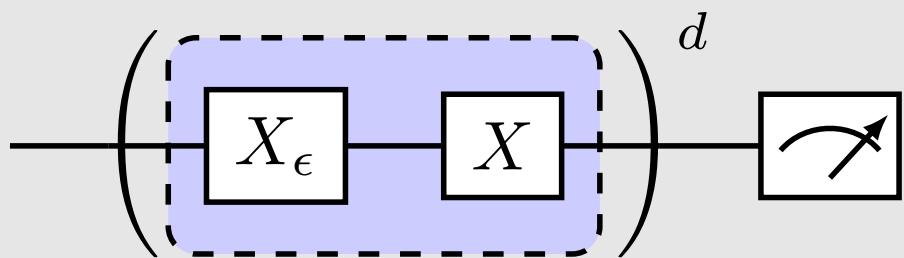
<sup>1</sup>Markov's inequality bounds the first moment of random variable. Use it when a constant probability bound is sufficient [1, Sec. 3.3].

<sup>2</sup>Chebyshev is derived from Markov. It bounds the second moment. It is the appropriate one when the variance  $\sigma$  is known. If  $\sigma$  is unknown, we can use the bounds of  $X \in [a, b]$ .

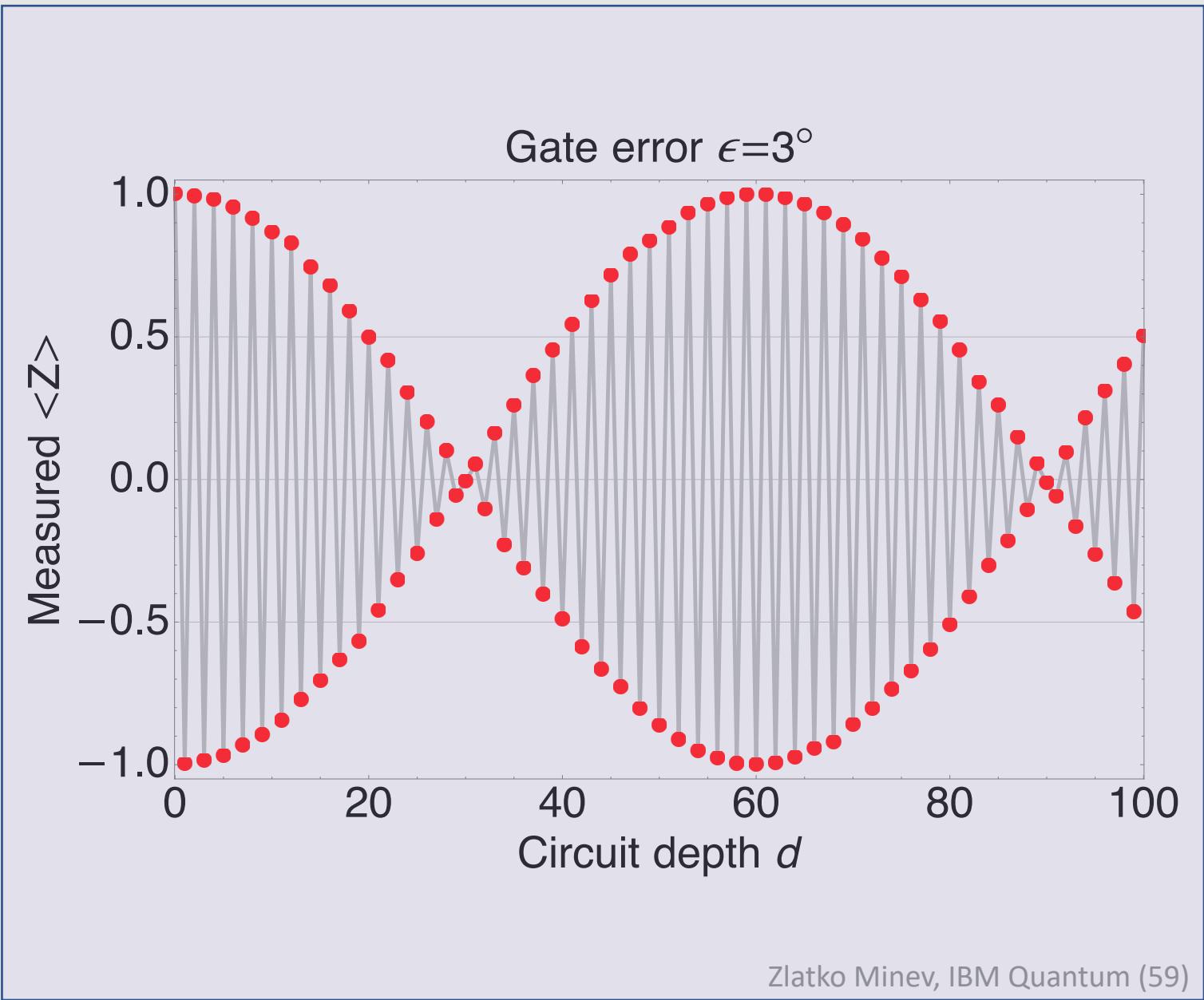
<sup>3</sup>Chernoff bound is used to bound the tails of the distribution for a sum of independent random variables. By far the most useful tool in randomized algorithms [1, Sec. 3.3].

<sup>4</sup>This probability can be interpreted as the level of significance  $\epsilon$  (probability of making an error) for a confidence interval around the mean of size  $2\epsilon$ . Therefore, we require at least  $\log(2\alpha)/2t^2$  samples to acquire  $1 - \alpha$  confidence interval  $\mathbb{E}[X] \pm t$ .

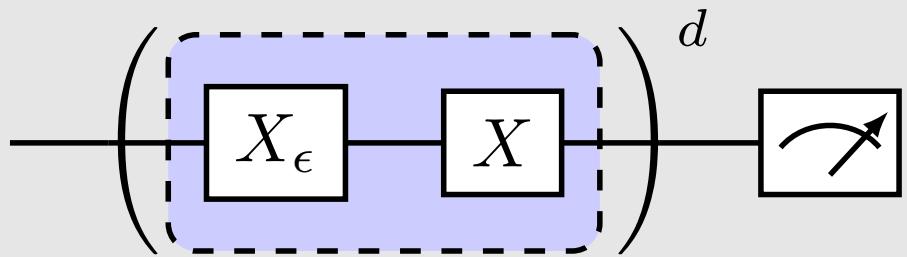
# Recall gate error result



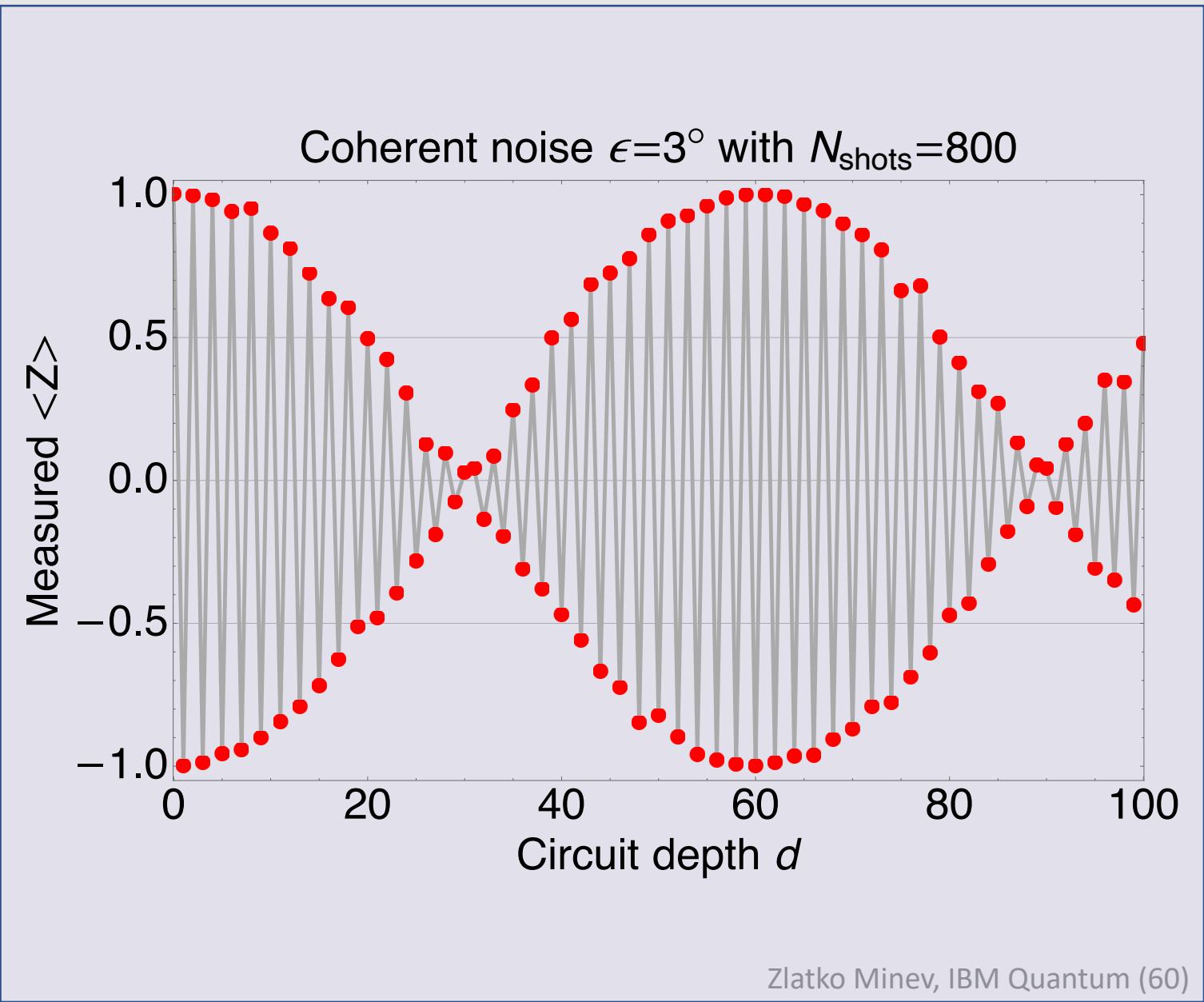
$$\langle \tilde{\psi}_f | Z | \tilde{\psi}_f \rangle = \cos(d\pi + d\epsilon)$$

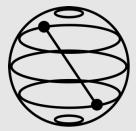


# Projection & sampling noise



$$\langle \tilde{\psi}_f | Z | \tilde{\psi}_f \rangle = \cos(d\pi + d\epsilon)$$





# Questions

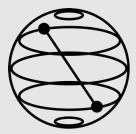
Answer these multiple-choice questions  
in the chat; for example, type “1a 2b.”

1. Projection noise is due to

- a) measurement apparatus that could be made more efficient
- b) classical limitations
- c) core nature of quantum physics

2. To reduce projection noise

- a) increase the number of sample
- b) you cannot undo it
- c) apply readout error mitigation



# Dive deeper? Try the following



1. Calculate the following for a qubit

1. The expectation value of the sample variance for N shots of the observable  $|1\rangle\langle 1|$ .

where the sample mean is defined as

$$S = \frac{1}{N} \sum_{n=1}^N M_n$$

and the sample variance is defined as

$$V = \frac{1}{N} \sum_{n=1}^N (M_n - S)^2$$

2. Is the estimate biased?

3. The variance of V.

4. Can you find an expression for an unbiased estimate of the sample variance?

2. What about two qubits?

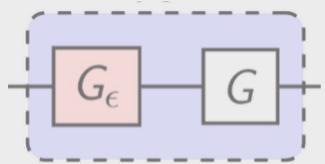
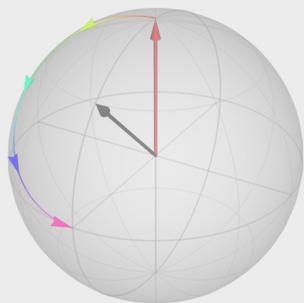
1. Can you find the projection operators for the observable ZZ?

2. Find the probability distribution for the observables ZI, IZ, and ZZ for a general state.

3. If you take 10 shots and find all 10 outcomes to be 1, what is the probability the qubit is in the  $|0\rangle$  state? (hint: it's not zero!)

# State preparation & measurement

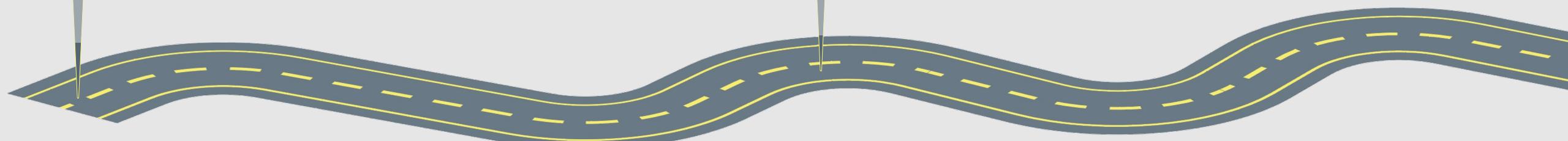
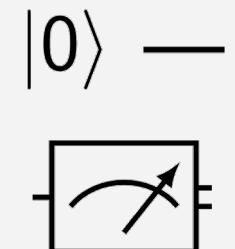
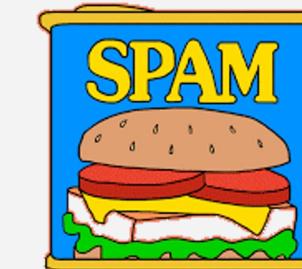
Coherent



Projection &  
measurement theory



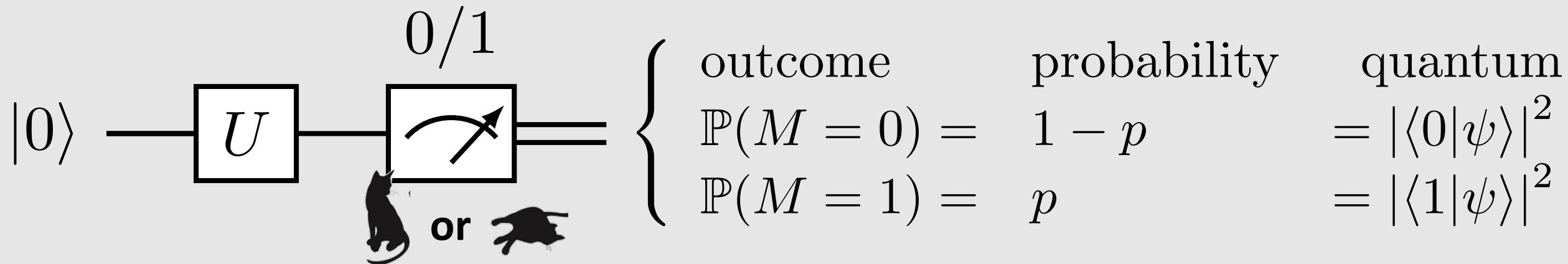
State preparation  
& measurement



coin toss: flaticon; spam: make it move;  
road based on: freepik

Zlatko Minev, IBM Quantum (63)

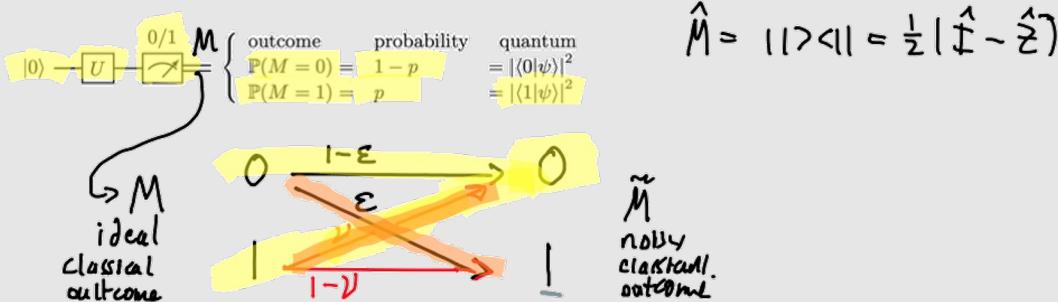
# Qubit example



## Measurement error

Qiskit Global Summer School on Quantum Machine Learning

Zlatko K. Minev



$$\hat{M} = |1\rangle\langle 1| = \frac{1}{2}(\hat{I} - \hat{Z})$$

$$P_M = \begin{pmatrix} P(M=0) \\ P(M=1) \end{pmatrix} = \begin{pmatrix} 1-p \\ p \end{pmatrix} \quad P_{\tilde{M}} = \begin{pmatrix} P(\tilde{M}=0) \\ P(\tilde{M}=1) \end{pmatrix} = \begin{pmatrix} 1-\tilde{p} \\ \tilde{p} \end{pmatrix}$$

$$\begin{cases} P(\tilde{M}=0) = P(\tilde{M}=0|M=0)P(M=0) + P(\tilde{M}=0|M=1)P(M=1) \\ P(\tilde{M}=1) = P(\tilde{M}=1|M=0)P(M=0) + P(\tilde{M}=1|M=1)P(M=1) = \varepsilon(1-p) + (1-\varepsilon)p = \tilde{p} \end{cases}$$

$$P_{\tilde{M}} = A P_M$$

$$A = \begin{matrix} & M=0 & M=1 \\ \tilde{M}=0 & \begin{pmatrix} P(\tilde{M}=0|M=0) & P(\tilde{M}=0|M=1) \\ P(\tilde{M}=1|M=0) & P(\tilde{M}=1|M=1) \end{pmatrix} & \\ \tilde{M}=1 & & \end{matrix}$$

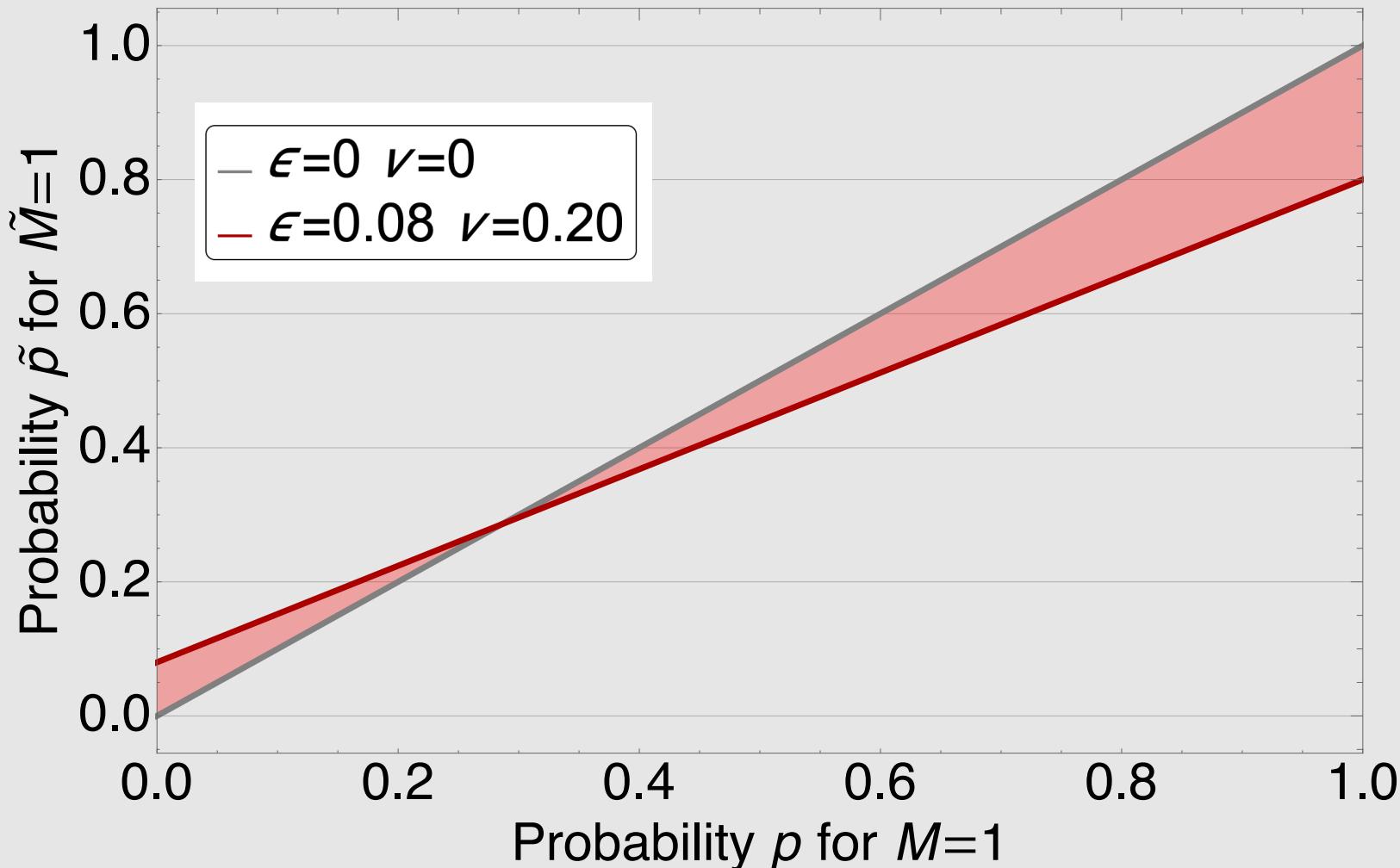
$$\approx \begin{pmatrix} 1-\varepsilon & \nu \\ \varepsilon & 1-\nu \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \text{ for an ideal measurement}$$

$$\sum_n A_{nn} = 1 \text{ for any } n \quad \text{Stochastic matrix}$$

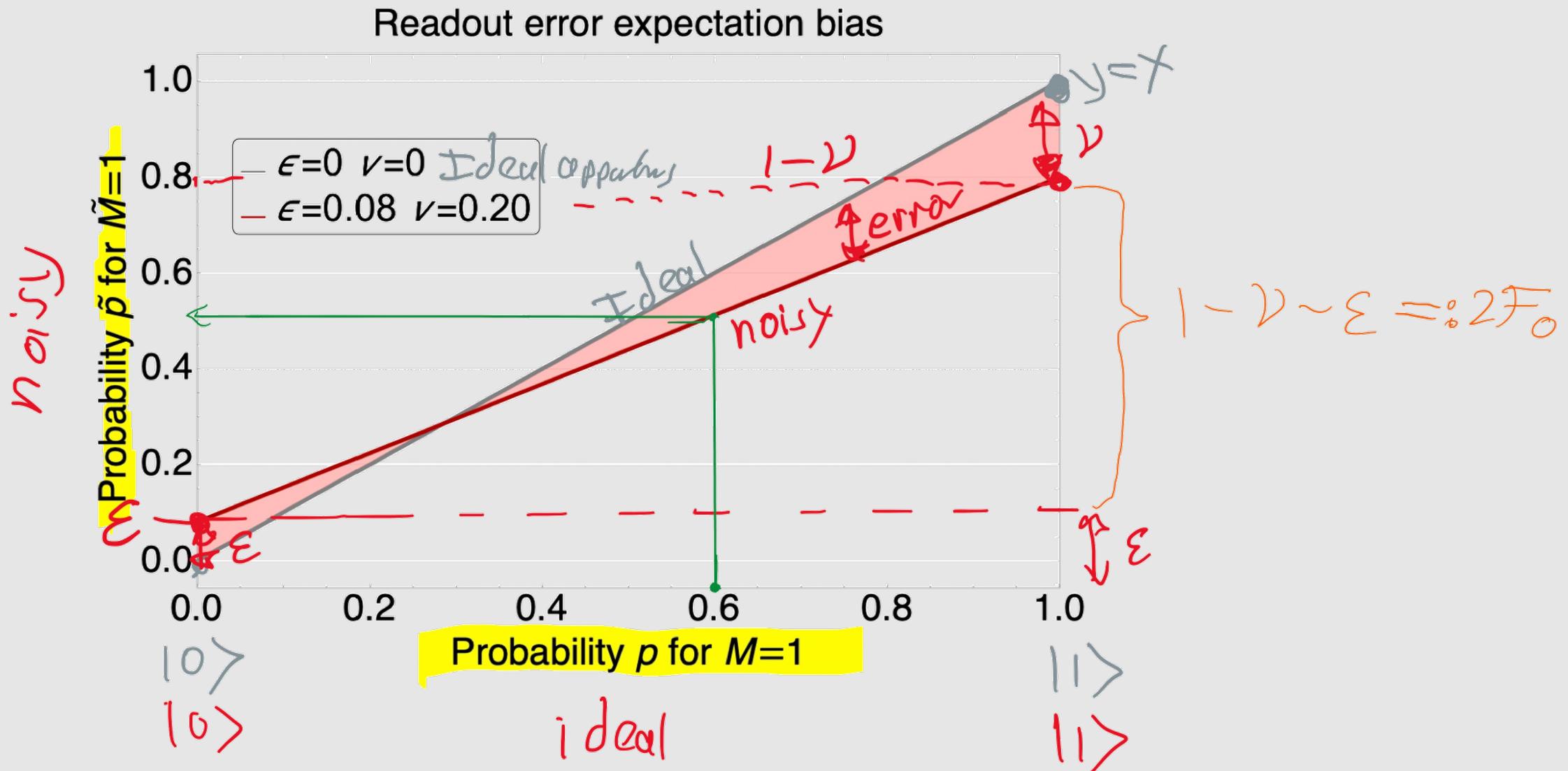
$$\begin{aligned} \tilde{p} &= \varepsilon(1-p) + (1-\varepsilon)p \\ &= \varepsilon - p\varepsilon + p - \nu p \\ &= p + \varepsilon - (\nu + \varepsilon)p \end{aligned}$$

# Readout error

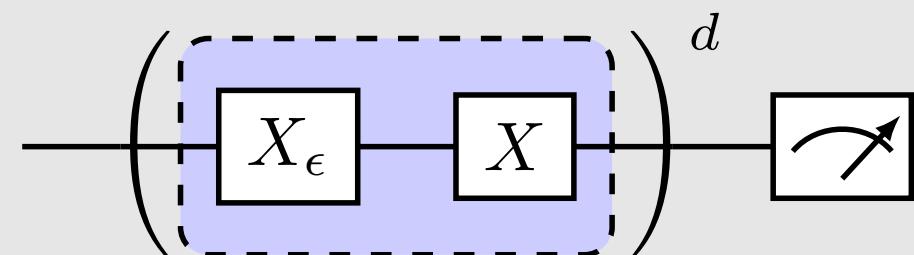
Readout error expectation bias

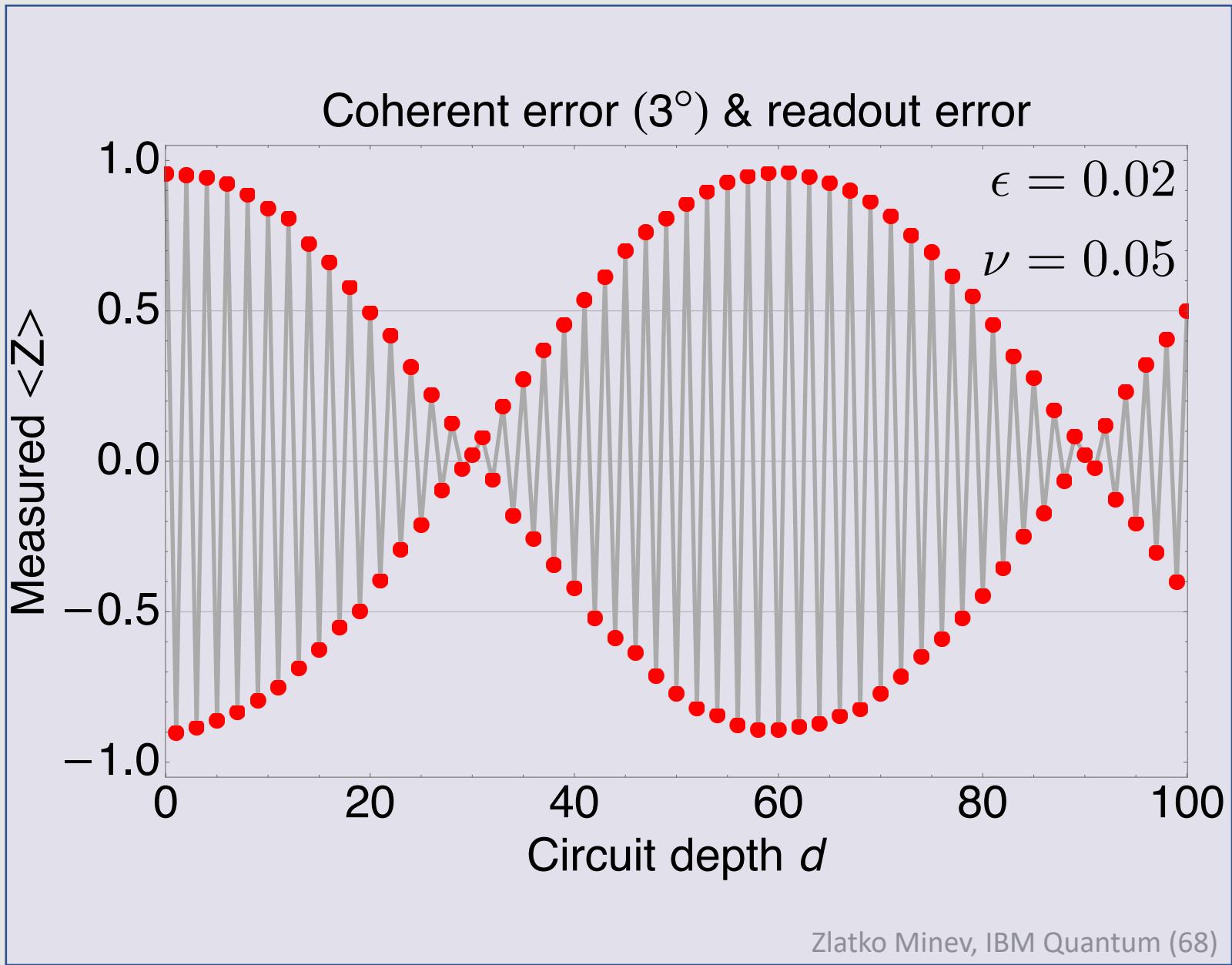


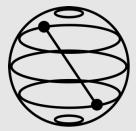
# Assignment fidelity



# Projection & sampling noise

$$A = \begin{matrix} M=0 & M=1 \\ \tilde{M}=0 & \begin{pmatrix} 1-\epsilon & \nu \\ \epsilon & 1-\nu \end{pmatrix} \\ \tilde{M}=1 \end{matrix}$$






# Questions

Answer these multiple-choice questions  
in the chat; for example, type “1a 2b.”

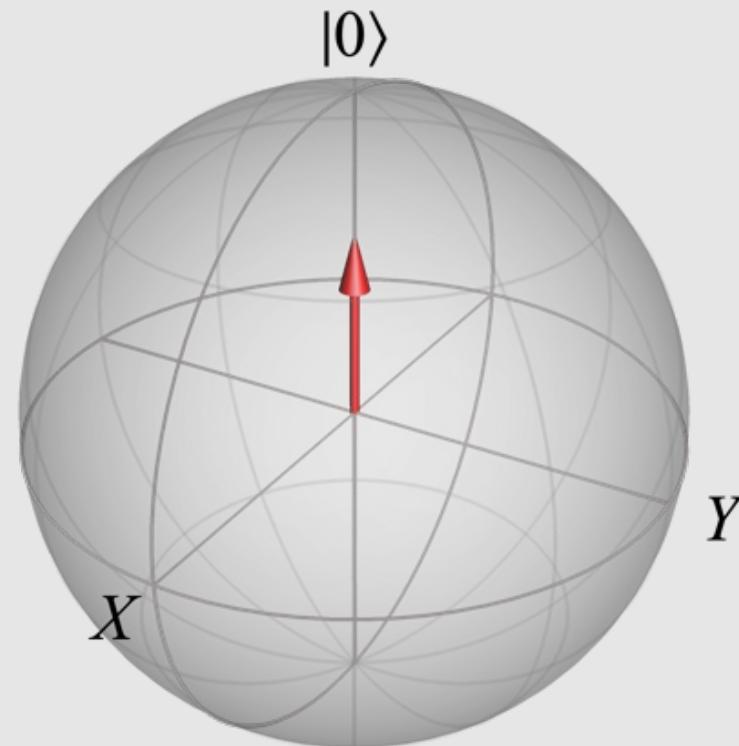
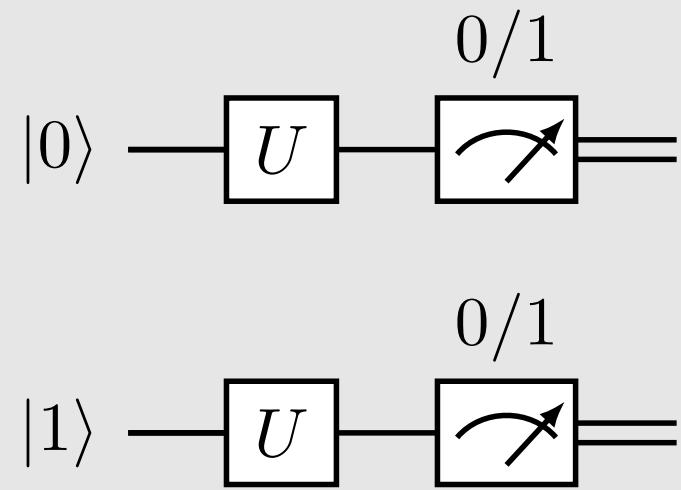
1. Readout error is due to

- a) measurement apparatus that could be made more efficient
- b) classical limitations
- c) core nature of quantum physics

2. To reduce readout error bias

- a) increase the number of sample
- b) you cannot undo it
- c) apply readout error mitigation

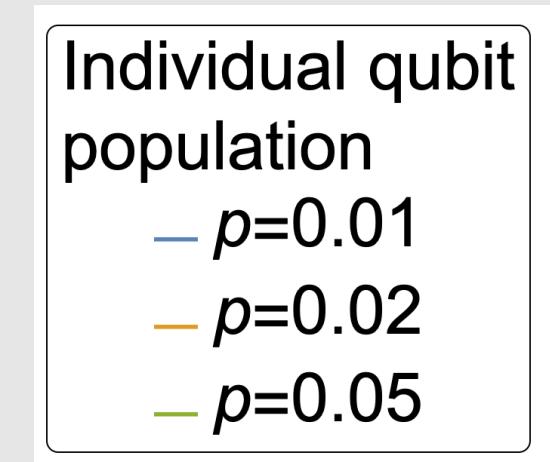
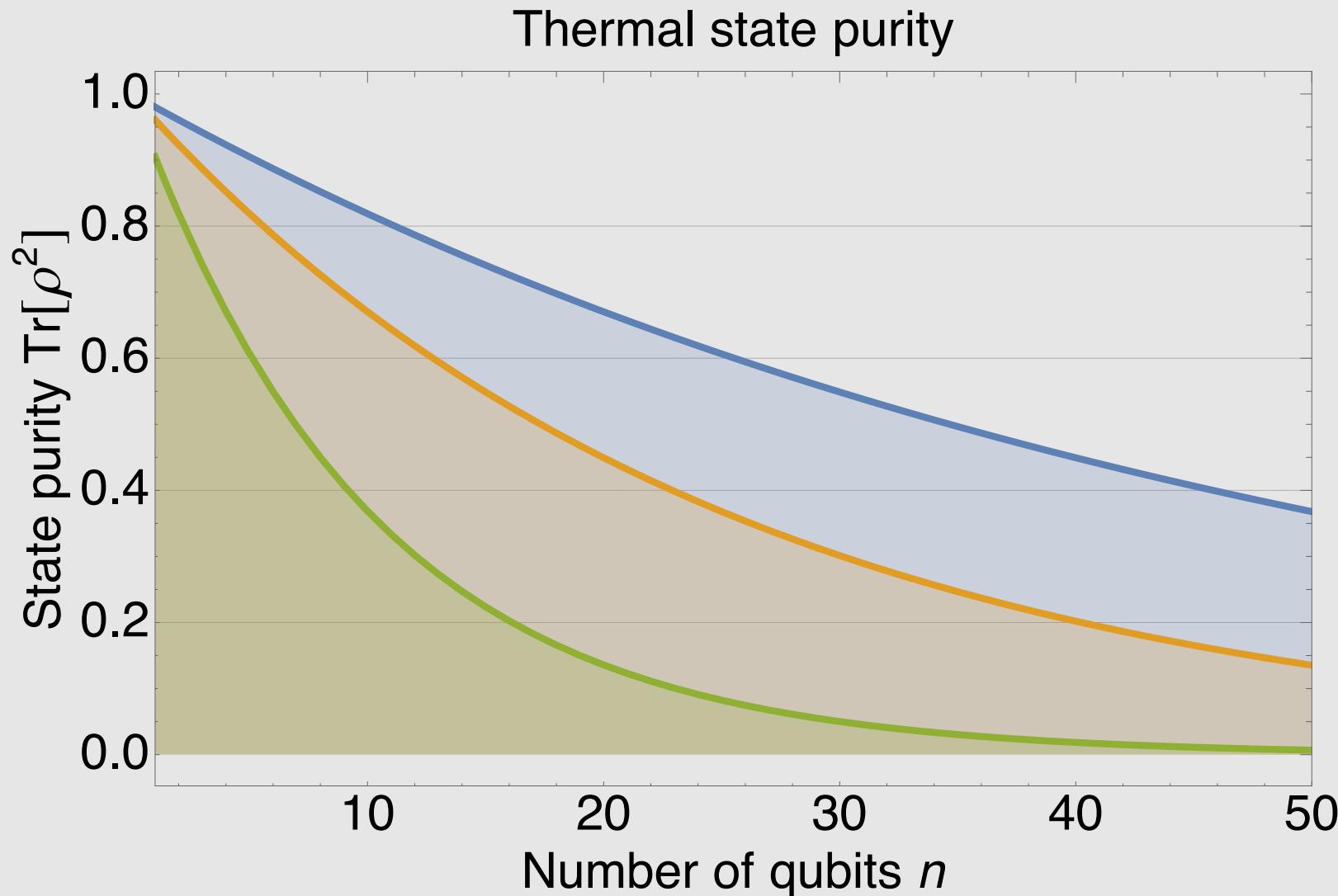
# State prep



# Multiple qubits

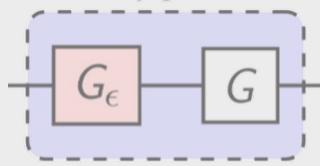
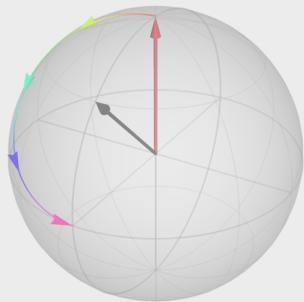
$$[(1 - p) |0\rangle\langle 0| + p |1\rangle\langle 1|]^{\otimes n} \equiv \boxed{U} \equiv \boxed{\curvearrowright}$$

# Thermal state

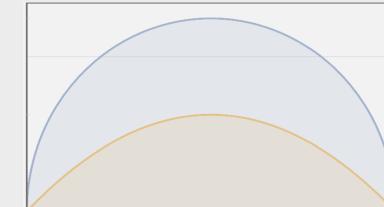


# Incoherent noise

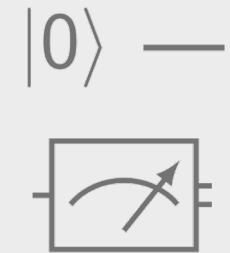
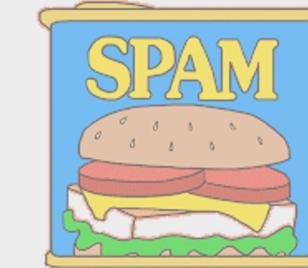
Coherent



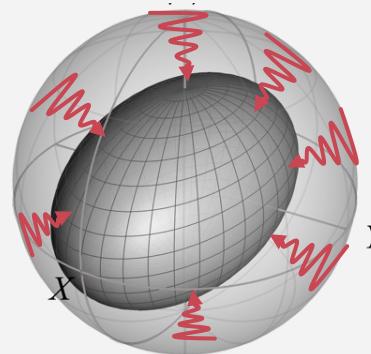
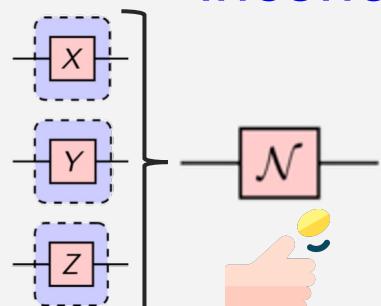
Projection & measurement theory



State preparation & measurement

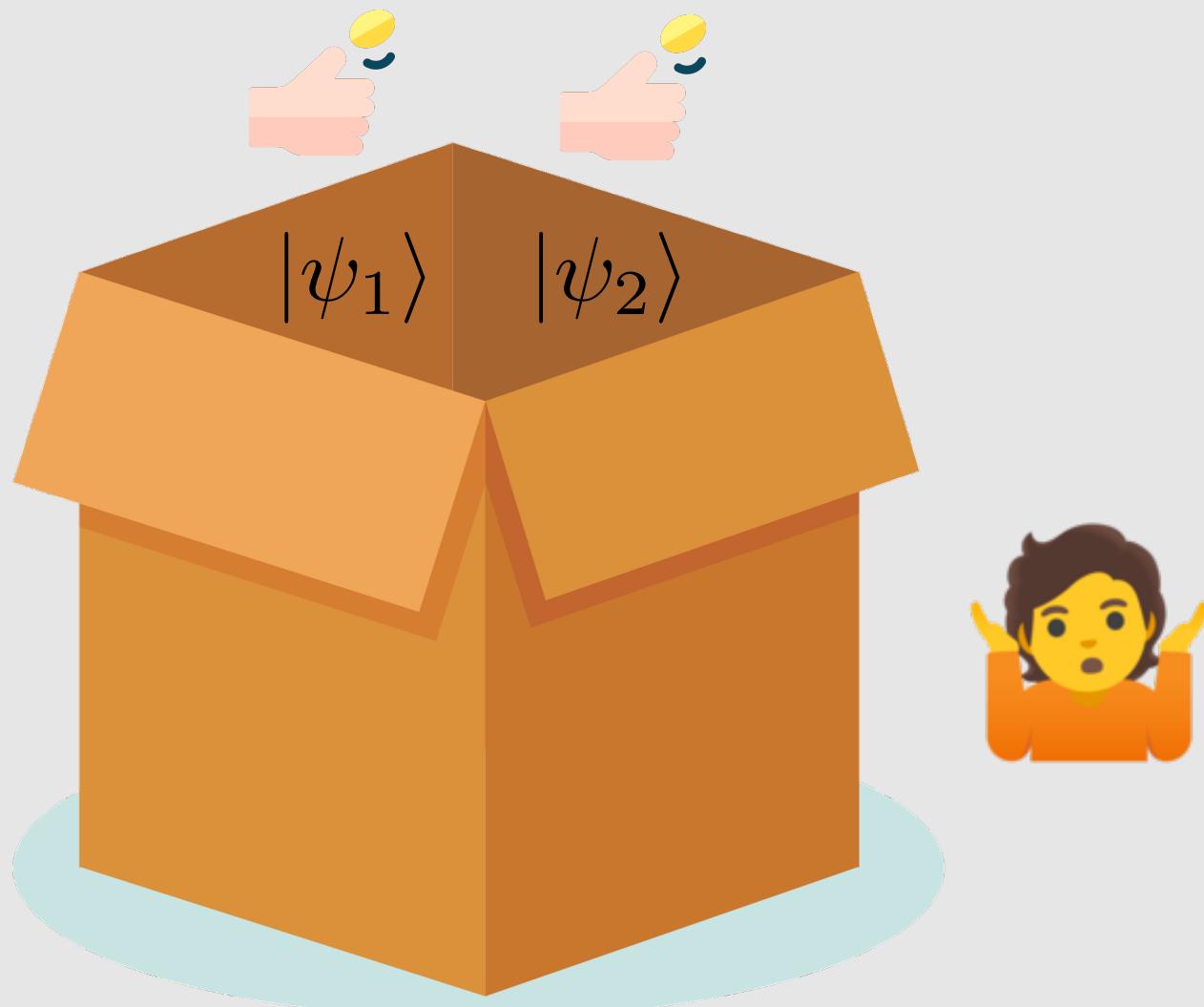


Incoherent noise

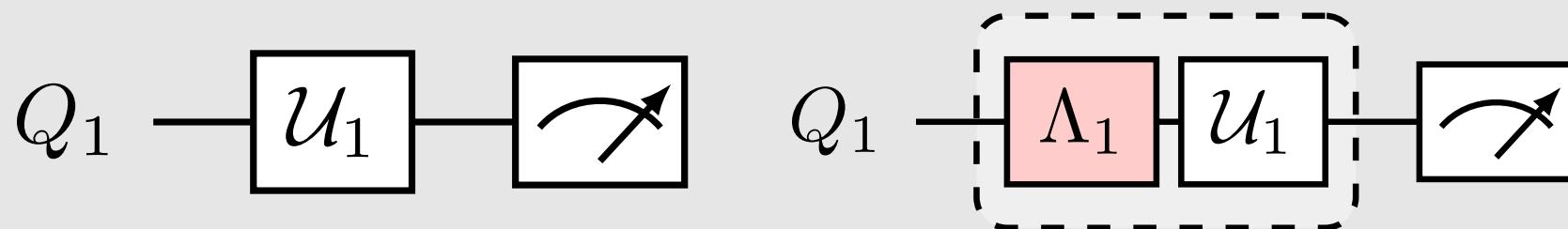


coin toss: flaticon; spam: make it move;  
road based on: freepik

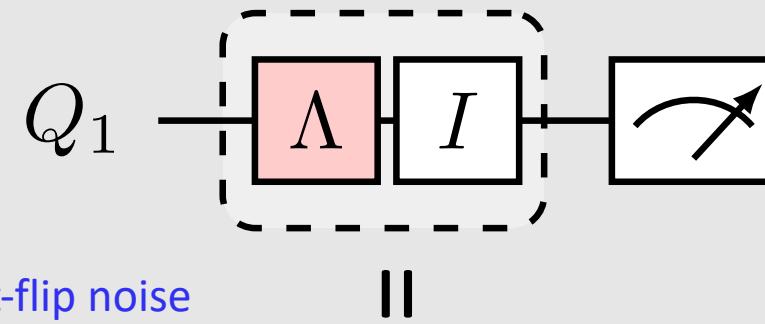
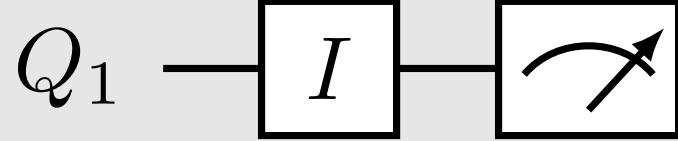
# Review: mixed state (density matrix)



# Toy model

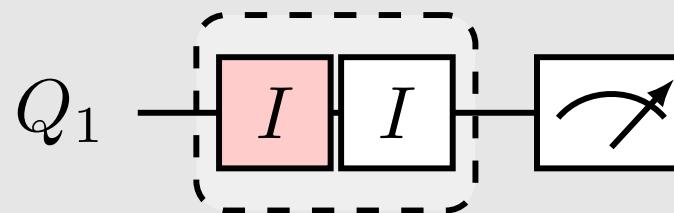
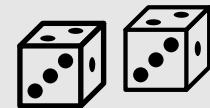


# Toy model: noise unraveling into quantum trajectories

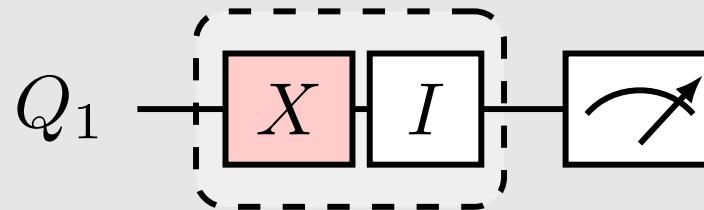


unraveling  
(quantum trajectories)

probability  $1-p$

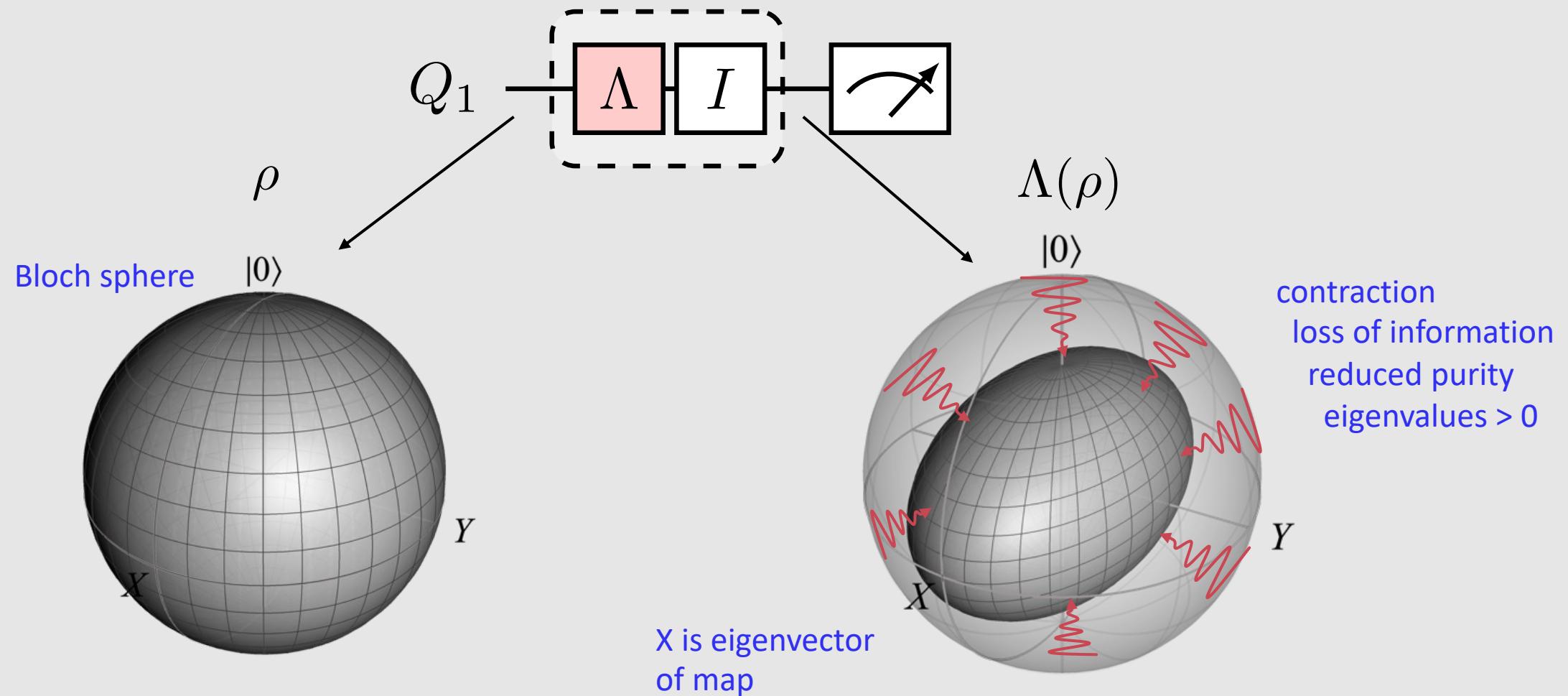


probability  $p$

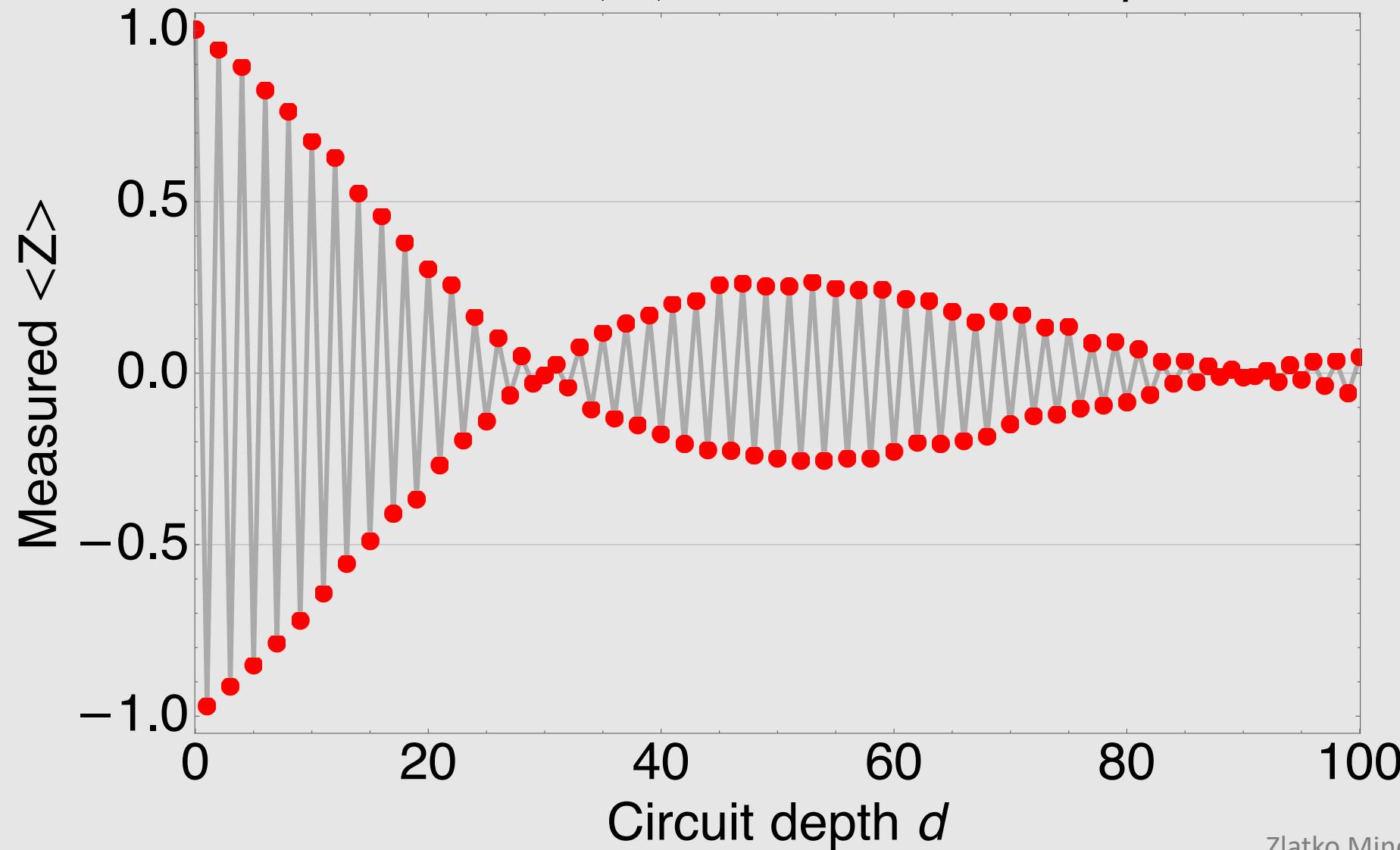


$$\Lambda(\rho) = (1 - p)I\rho I + pX\rho X$$

# Toy model: noise unraveling into quantum trajectories



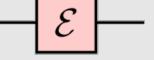
## Coherent error ( $3^\circ$ ) & incoherent error $p=0.012$



X bit-flip noise



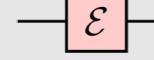
Phase noise



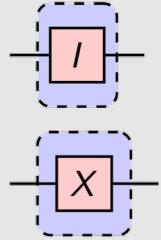
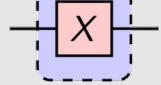
Depolarizing noise



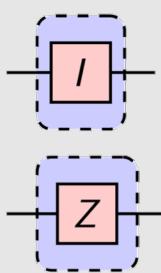
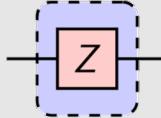
Stochastic Pauli noise



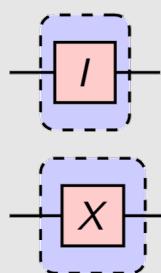
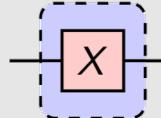
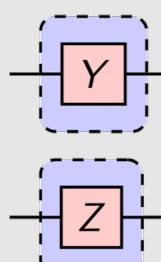
probability circuit instance

 $1 - p$  $p$ 

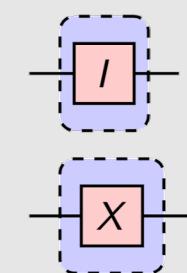
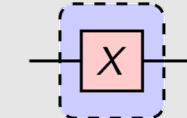
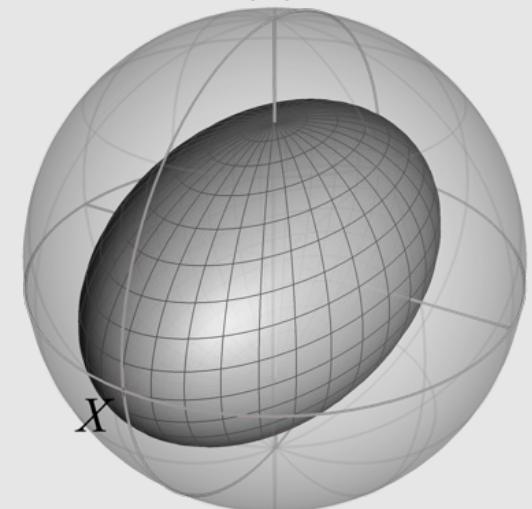
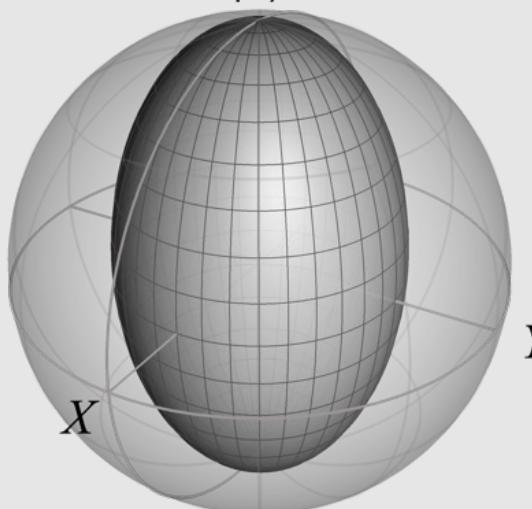
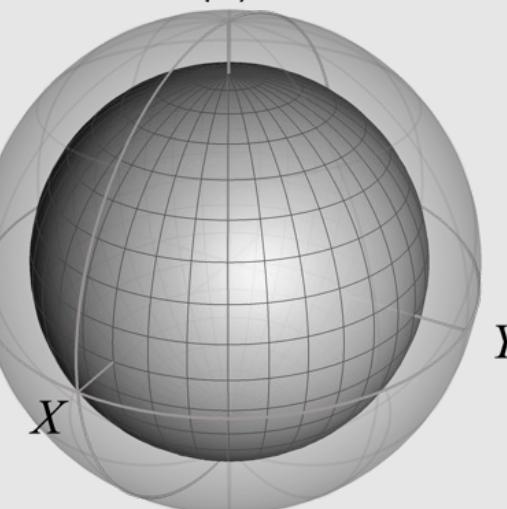
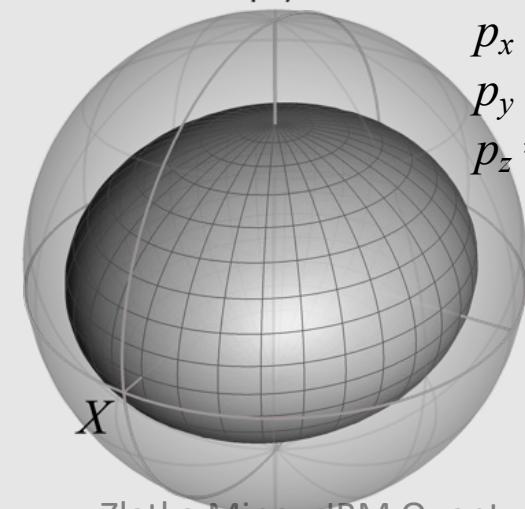
probability circuit instance

 $1 - p$  $p$ 

probability circuit instance

 $1 - 3p/4$  $p/4$  $p/4$  $p/4$ 

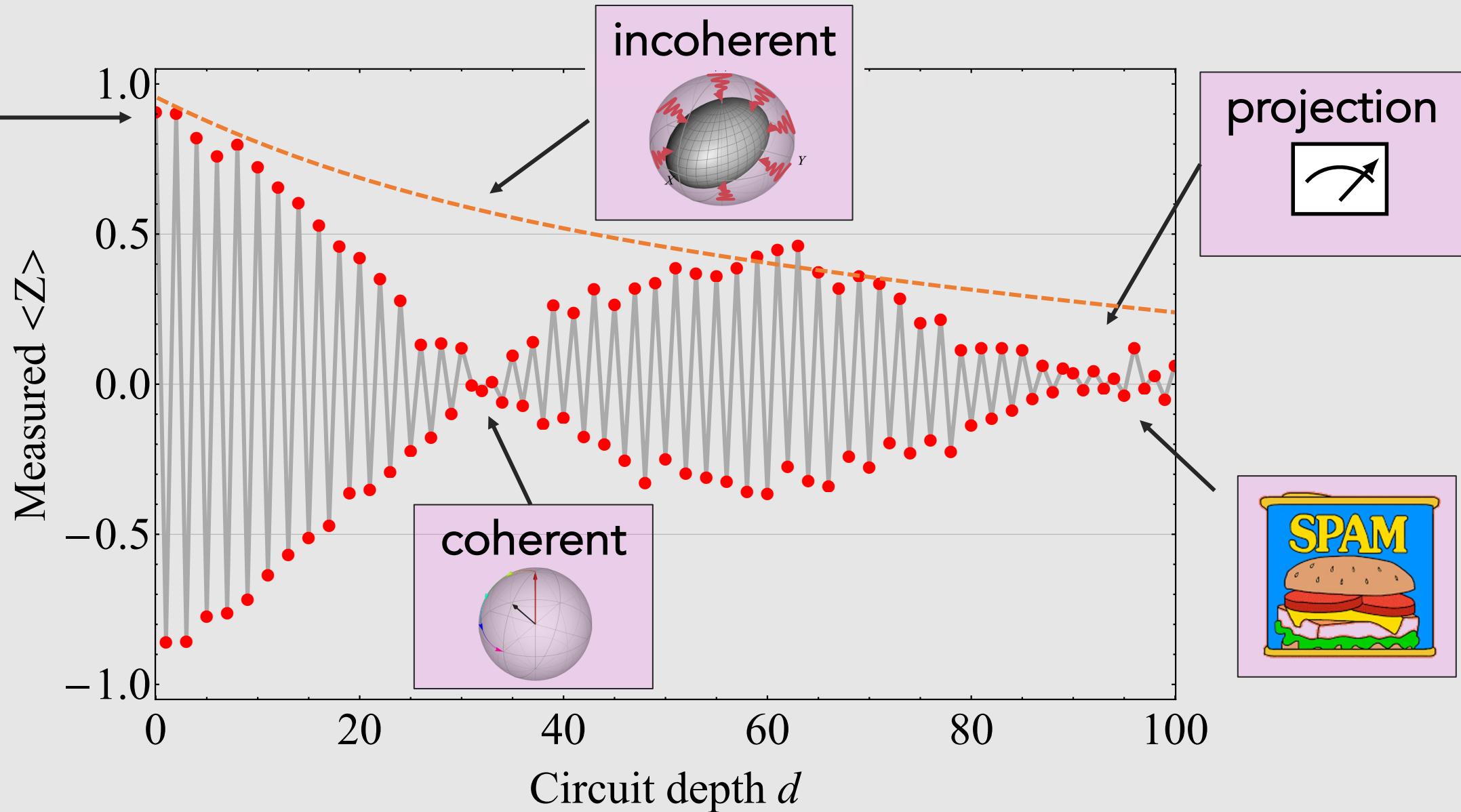
probability circuit instance

 $1 - p_X - p_Y - p_Z$  $p_X$  $p_Y$  $p_Z$  $|0\rangle$  $|0\rangle$  $|0\rangle$  $|0\rangle$ 

$$\begin{aligned} p_x &= 0.1 \\ p_y &= 0.2 \\ p_z &= 0.4 \end{aligned}$$

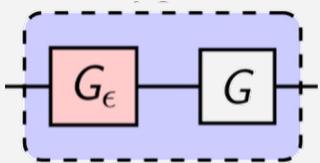
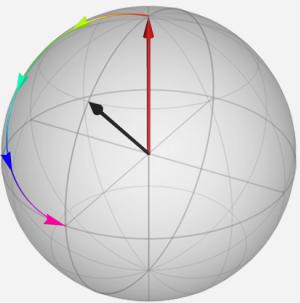


# Elements of noise

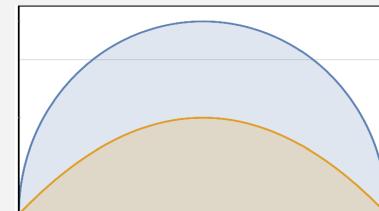
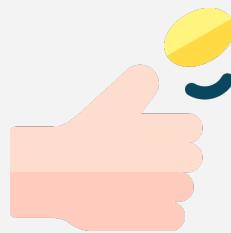


# Our journey so far

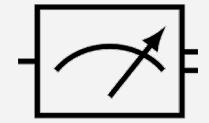
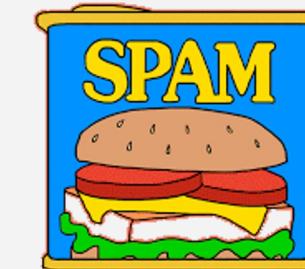
## Coherent noise



## Measurement in a nutshell Projection noise

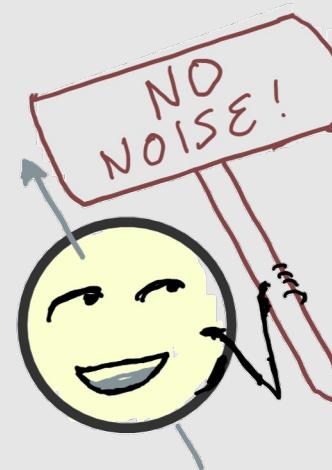
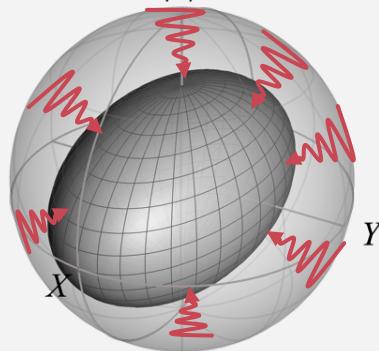
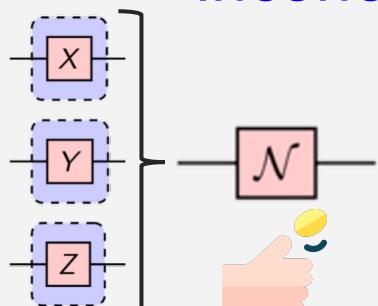


## SPAM: Noisy meters



$|0\rangle$  —

## Incoherent noise

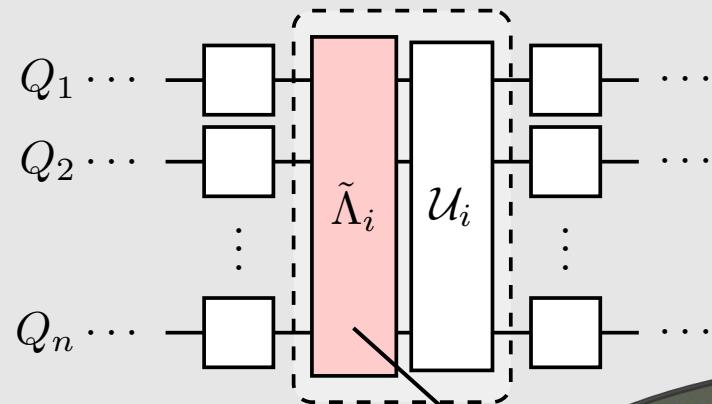


## Bonus content Coherent ZZ State preparation

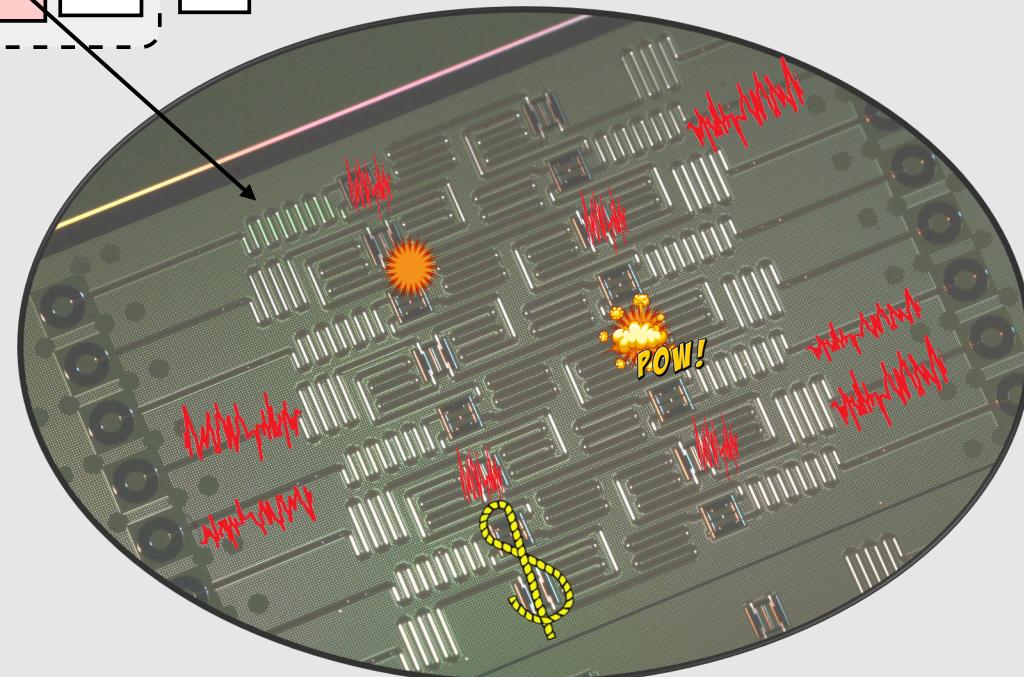


coin toss: flaticon; spam: make it move;  
road based on: freepik

# Outlook: Is it possible to learn & correct the noise with accuracy, efficiency, and scalability?

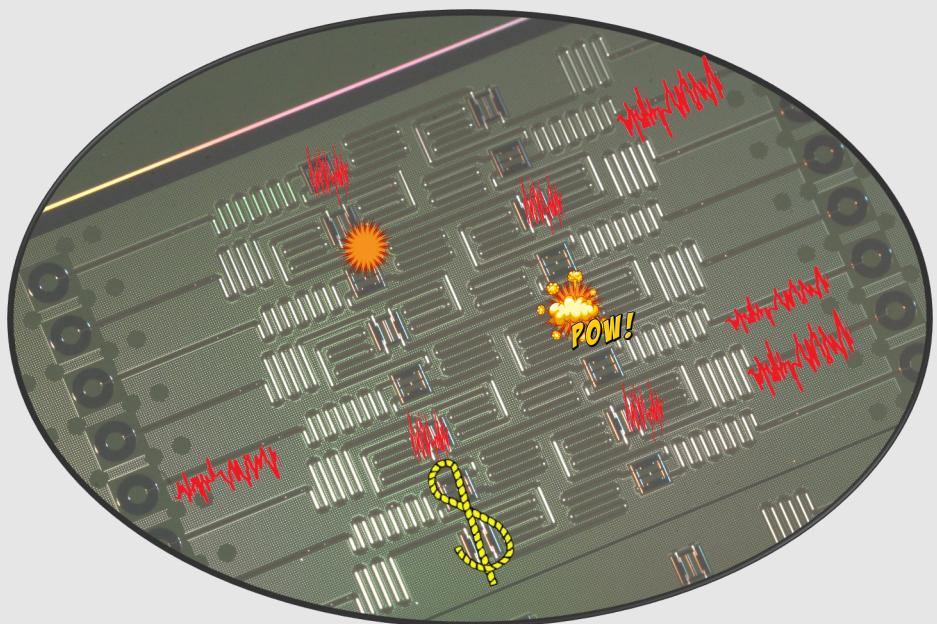


Energy relaxation  $T_1$   
Dephasing  $T_2$   
Coherent errors ZZ  
Classical crosstalk  
Quantum crosstalk  
State preparation error  
Measurement correlated errors  
...



Control errors  
Photon shot noise  
1/f charge noise  
1/f flux noise  
Nonequilibrium quasiparticles  
Leakage  
Cosmic rays  
...

# Error mitigation and error correction



## Error mitigation: working with what you have

- **benefit** suppress errors on classical results (expectation values)
- **q-cost** no extra qubits or hardware resources needed
- **c-cost** trades classical resources (post-processing) for lower error
- **limitation** bad asymptotic scaling: high number of samples & circuits

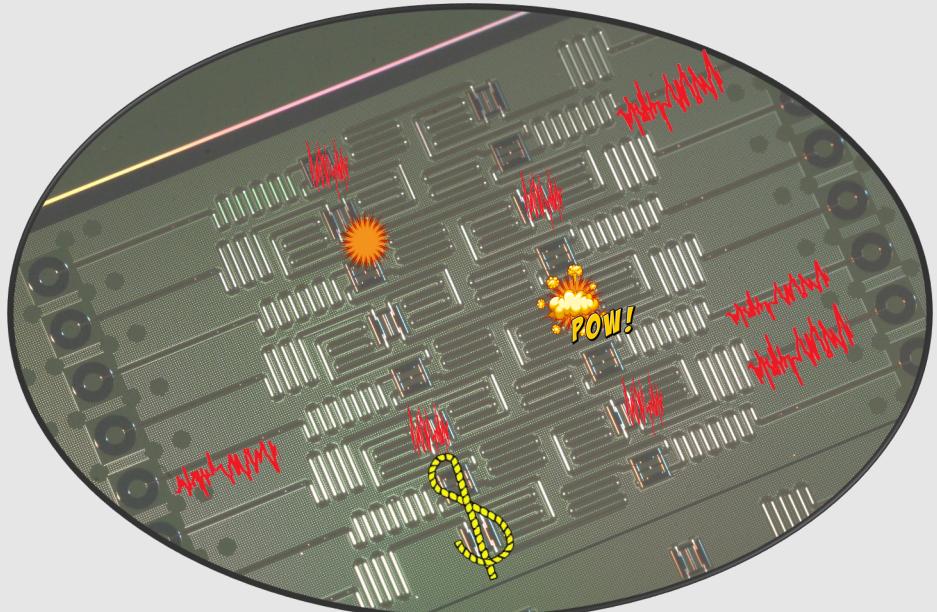
## Error correction: protecting quantum information

- **benefit** suppress & correct errors to arbitrarily small level
- **q-cost** very large qubit and hardware overhead
- **c-cost** decoding and encoding can be classically costly
- **challenge** requires fault-tolerant operations and readout

# Latest: Qiskit Quantum Seminar YouTube Series

Ways to learn more

TODO



# Introduction to Quantum Noise

Lab work with Qiskit

Run experiments on real devices

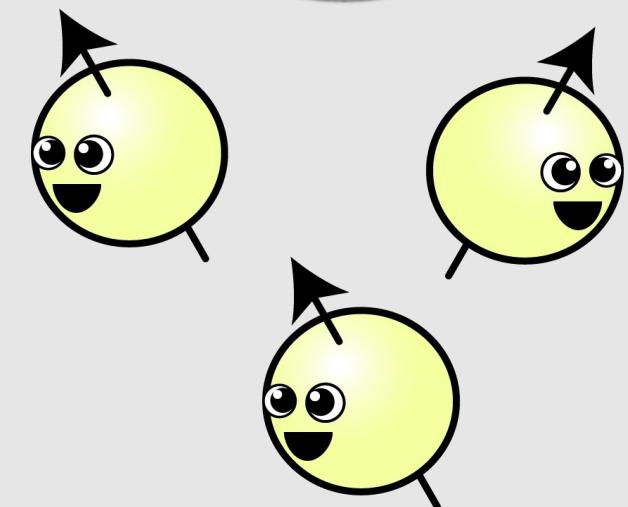
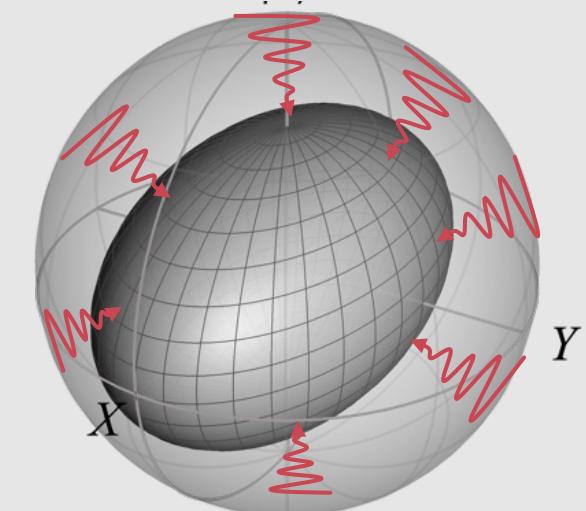


Check out references, problems given in  
the lecture, dangerous bends

Stay in touch

Thank you!

Zlatko K. Minev



@zlatko\_minev



zlatko-minev.com

IBM Quantum

The important thing is not to stop questioning.  
Curiosity has its own reason for existence.

One cannot help but be in awe when he  
contemplates the mysteries of eternity, of life, of the  
marvelous structure of reality.

It is enough if one tries merely to comprehend a  
little of this mystery each day.

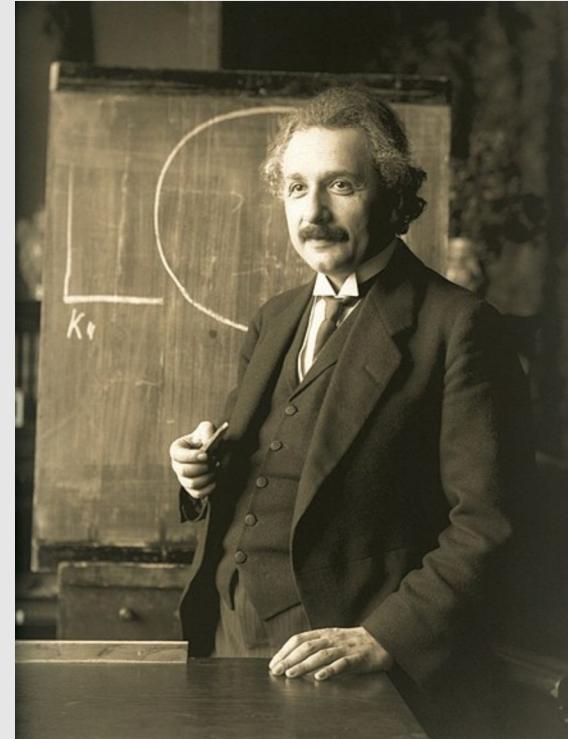


Photo: F. Schmutz

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Albert Einstein



@zlatko\_minev



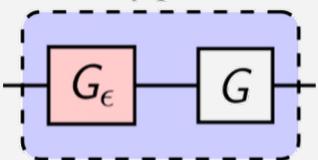
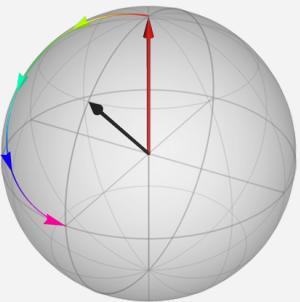
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IBM Quantum

# Bonus content

# Bonus content

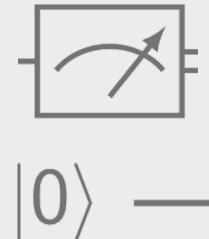
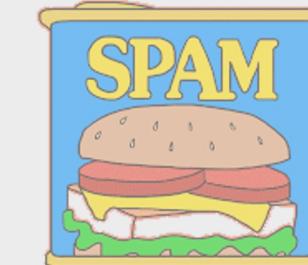
## Coherent noise



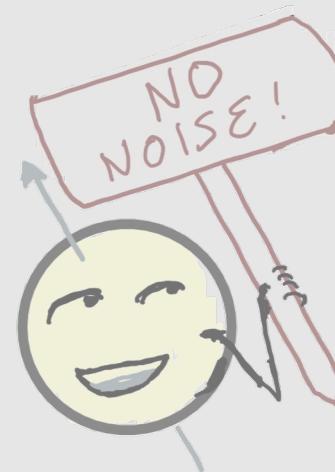
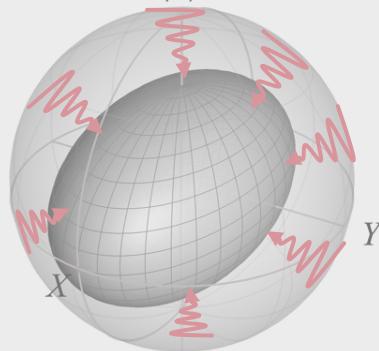
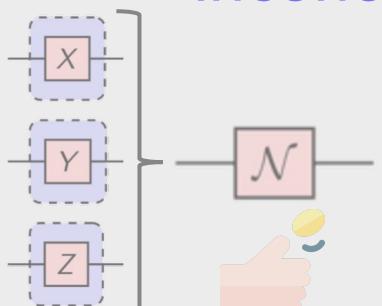
## Measurement in a nutshell Projection noise



## SPAM: Noisy meters



## Incoherent noise

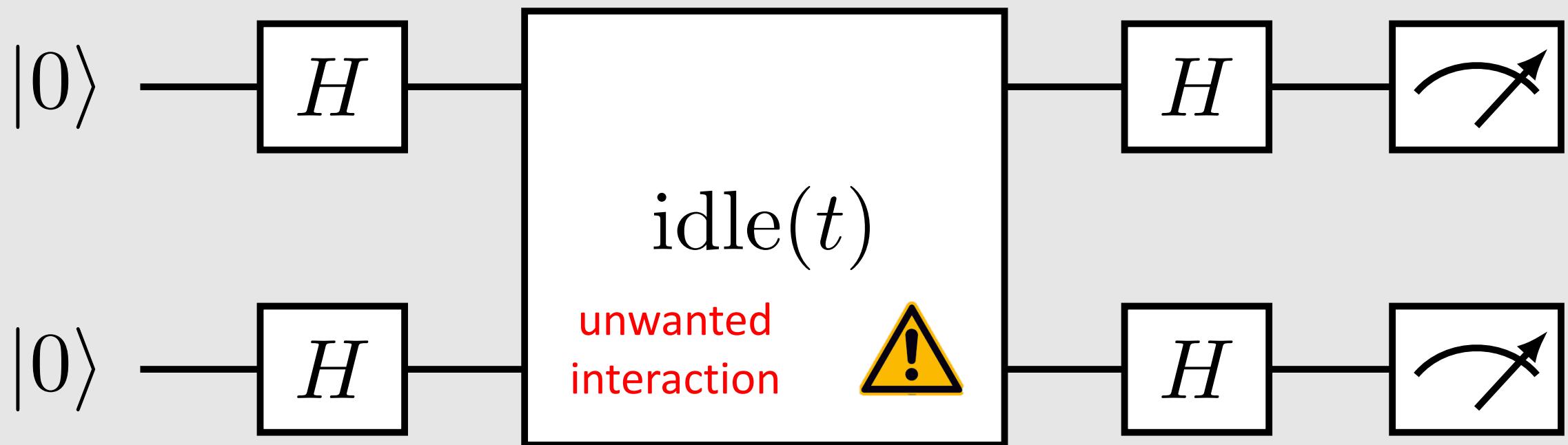


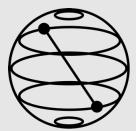
## Bonus content



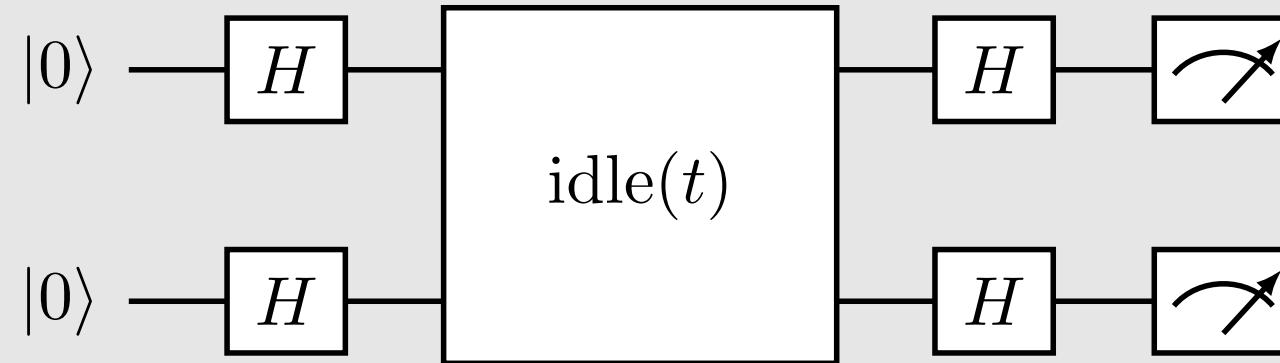
coin toss: flaticon; spam: make it move;  
road based on: freepik

## Bonus content: two-qubit coherent ZZ error





# Bonus content: two-qubit ZZ error



Hadamard gate



$$H = \begin{matrix} |0\rangle & \langle 0| \\ |1\rangle & \langle 1| \end{matrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{pmatrix}$$

$$H |0\rangle = |+x\rangle = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$H |1\rangle = |-x\rangle = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

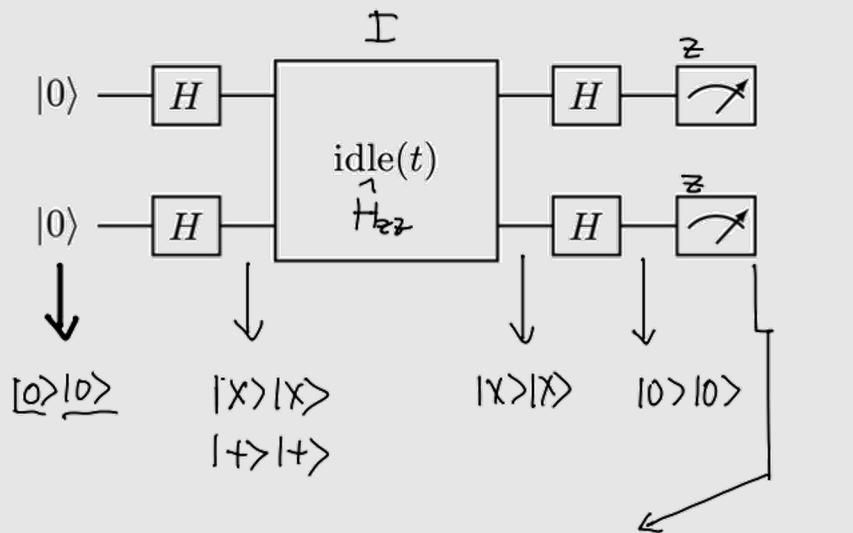
Breakout to notebook

# Introduction to quantum noise

## Coherent errors

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$$\begin{aligned}\langle z \rangle &= + \\ \langle 1 \rangle &= + \\ \langle z z \rangle &= \langle 0 | \cancel{z} | \cancel{z} | 0 \rangle \\ &= \langle 0 | z | 0 \rangle \langle 0 | z | 0 \rangle \\ &= (+)(+) \\ &\approx +\end{aligned}$$

Hadamard gate

$$H = \begin{bmatrix} |0\rangle & \langle 0| \\ |1\rangle & \langle 1| \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

$$\begin{cases} H|0\rangle = |+x\rangle = \frac{1}{\sqrt{2}}(1) \\ H|1\rangle = |-x\rangle = \frac{1}{\sqrt{2}}(1) \end{cases}$$

$$|+x\rangle = +|+x\rangle$$

$$|-x\rangle = -|-x\rangle$$

$$|+x\rangle := |+\rangle$$

$$|-x\rangle := |-\rangle$$

$$\begin{cases} z|0\rangle = +|0\rangle \\ z|1\rangle = -|1\rangle \end{cases}$$

Now

 $zz$  Interaction

$$\hat{H} = \frac{1}{2} \hbar \omega \hat{Z}\hat{Z}$$

$$\begin{aligned}\hat{U}(t) &= \exp(-i\hbar^{-1} \hat{H} t) \\ &= \exp\left(-i \frac{\omega t}{2} \hat{Z}\hat{Z}\right) \\ &\approx \cos\left(\frac{\omega t}{2}\right) \hat{I} - i \sin\left(\frac{\omega t}{2}\right) \hat{Z}\hat{Z} \\ &= \hat{R}_{\hat{Z}\hat{Z}} (\theta = \omega t)\end{aligned}$$

$$\begin{aligned}\hat{R}_x(\theta) &= \exp\left(-i \frac{\theta}{2} \hat{X}\right) \quad \hat{X}^2 = \hat{I} \\ &= \cos\left(\frac{\theta}{2}\right) \hat{I} - i \sin\left(\frac{\theta}{2}\right) \hat{X} \\ (\hat{Z}\hat{Z})^2 &= Z^2 Z^2 = \hat{I}_2 \otimes \hat{I}_2 = \underline{\underline{I_4}}\end{aligned}$$

$$|0\rangle|0\rangle \xrightarrow{HH} |+\rangle|+\rangle \xrightarrow{R_{zz}(\theta)} |+\rangle|+\rangle = \cos\frac{\theta}{2} |+\rangle|+\rangle - i \sin\frac{\theta}{2} |-\rangle|-\rangle \xrightarrow{H+I} \underline{\underline{\cos\frac{\theta}{2} |0\rangle|0\rangle - i \sin\frac{\theta}{2} |1\rangle|1\rangle}}$$

$$R_{zz}(\theta)|+\rangle|+\rangle = \cos\frac{\theta}{2}|+\rangle|+\rangle - i \sin\frac{\theta}{2}(Z|+\rangle) \otimes (Z|+\rangle)$$

$$Z|+\rangle = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = |-\rangle$$

$$Z|+\rangle = |-\rangle$$

$$Z|-\rangle = |+\rangle$$

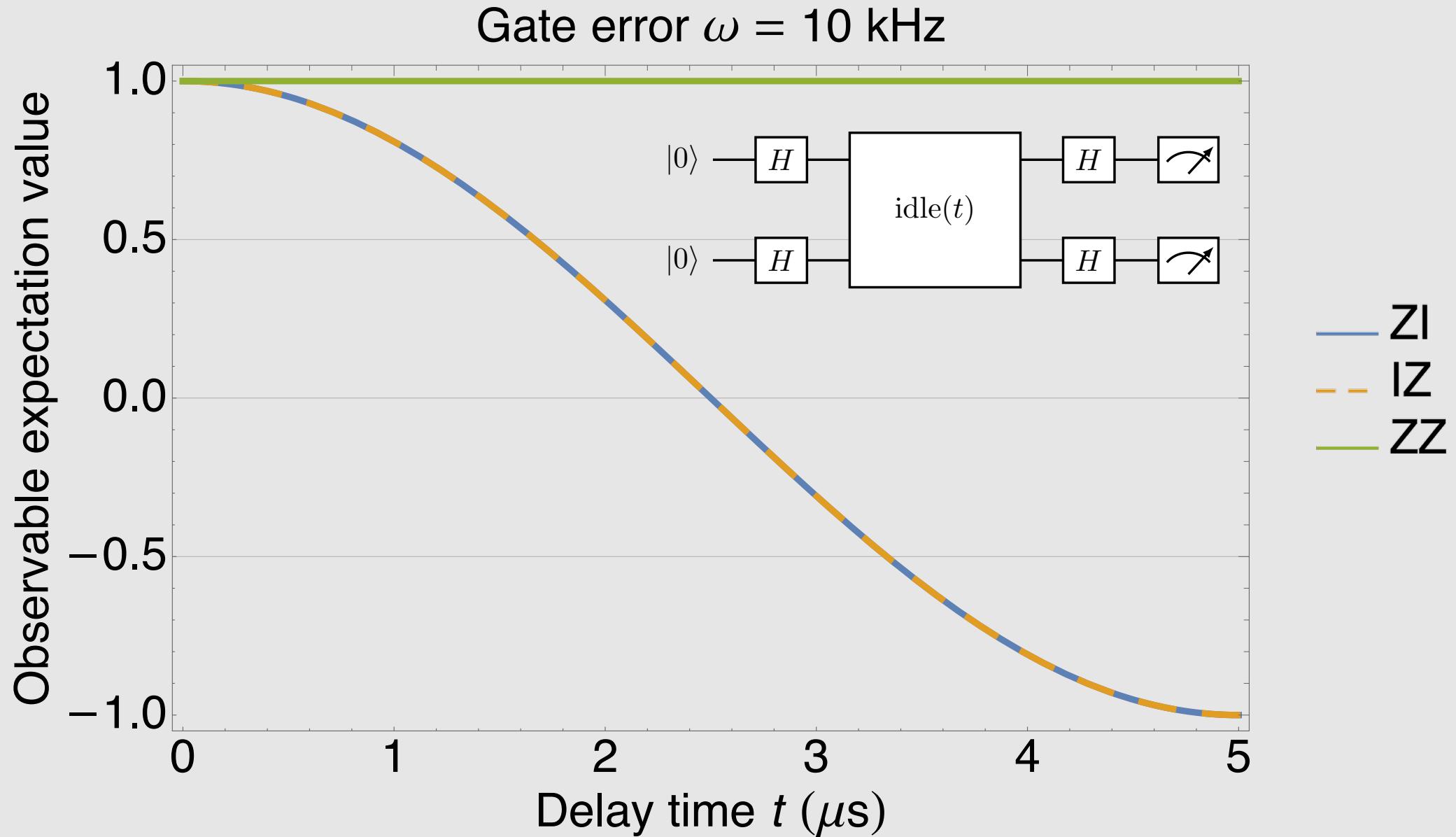
$$\langle ZI \rangle = \cos \omega t$$

$$\langle IZ \rangle = \cos \omega t$$

$$\langle ZZ \rangle = 1$$

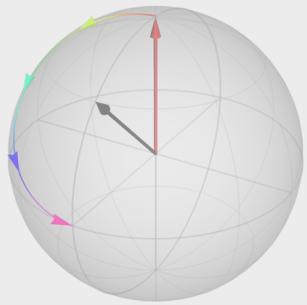


# Bonus content: two-qubit ZZ error



# Bonus content

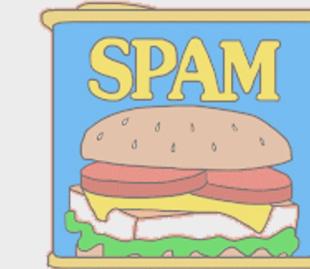
## Coherent noise



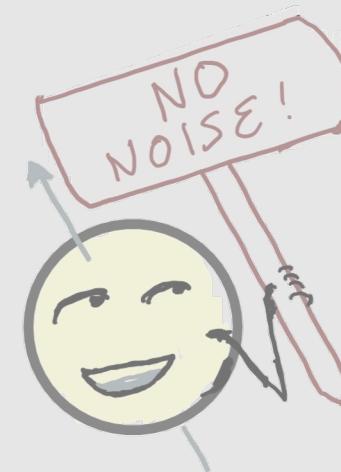
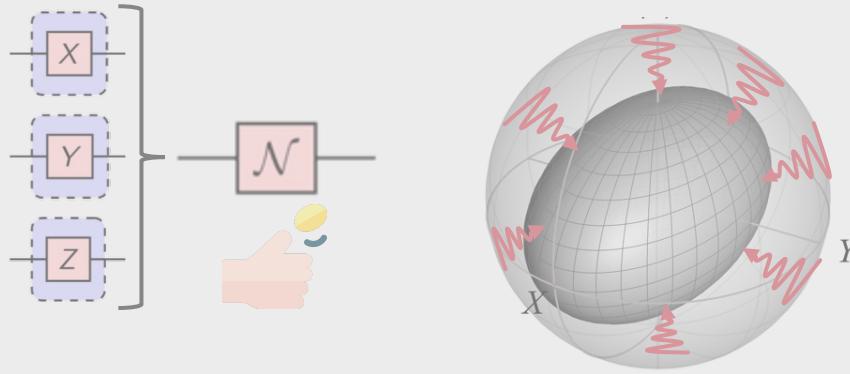
## Measurement in a nutshell Projection noise



## SPAM: Noisy meters



## Incoherent noise



## Bonus content



coin toss: flaticon; spam: make it move;  
road based on: freepik

# A matrix

Bonus section content:

Reconstruct A matrix

$$|0\rangle \xrightarrow{A} (\hat{M}=0) \xrightarrow{A} \tilde{M} \\ p=0 \qquad \qquad \hat{p}=\varepsilon$$

$$|0\rangle \xrightarrow{\text{X}} (\hat{M}=1) \xrightarrow{A} \tilde{M} \leftarrow w_{\text{access}}^{\text{banc}} \\ p=1 \qquad \qquad \hat{p}=1-\varepsilon$$

Noise mitigation

We know A

measure  $\hat{p}$ ,  $\tilde{P}_M$  noisy

find  $p$ ,  $P_M$  ideal

$$\dim A = 2^n \times 2^n \quad n = \text{#qubits}$$

$$\tilde{P}_M = A P_M$$

$$P_M = A^{-1} \tilde{P}_M$$

$$p =$$

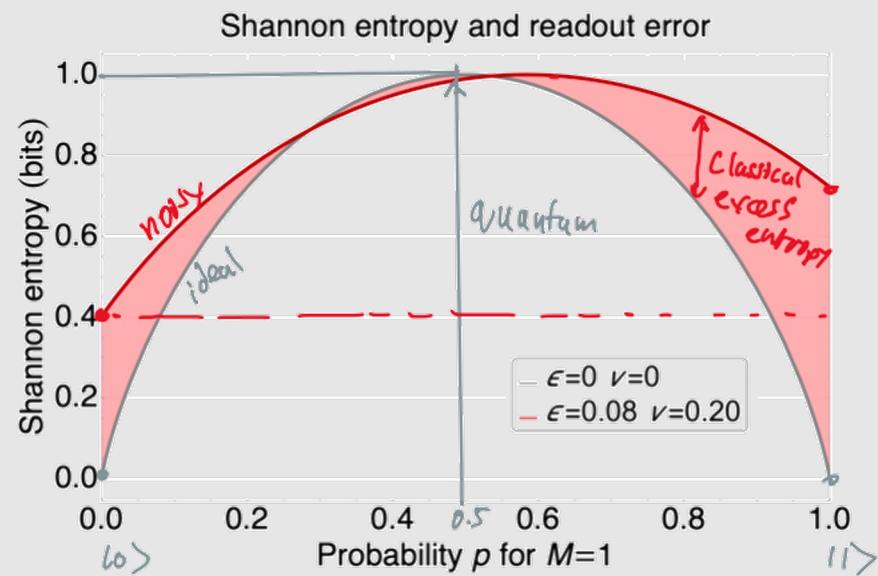
Assessment Fidelity

$$\begin{aligned} \hat{F}_0 &= 1 - \frac{1}{2} [p(\tilde{M}=1|M=0) + p(\tilde{M}=0|M=1)] \\ &= \frac{1}{d} \text{Tr}(A) \\ &= 1 - \frac{1}{2} (M+D) \end{aligned} \quad d = 2^n, n = \text{#qubits}$$

Shannon Entropy

$$H(A) = H(P_M) = - \sum_m P_m \log_2 P_m = - (1-p) \log_2 (1-p) - p \log_2 p$$

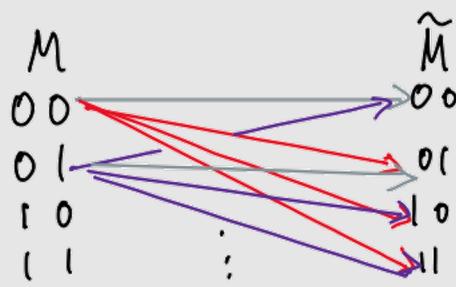
Binary entropy



Larger System

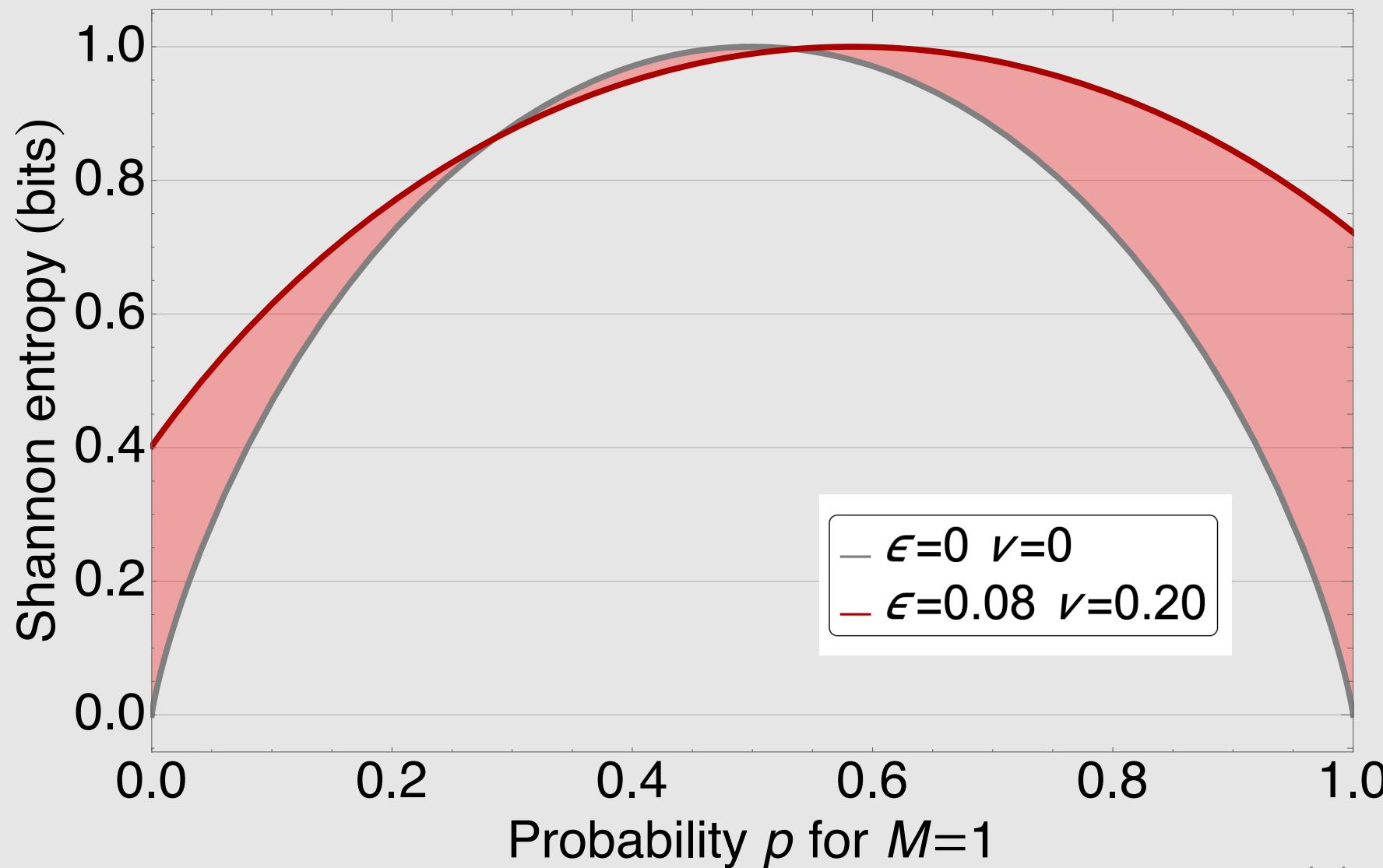
n qubits

$$\dim A = 2^n \times 2^n$$



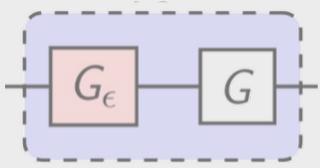
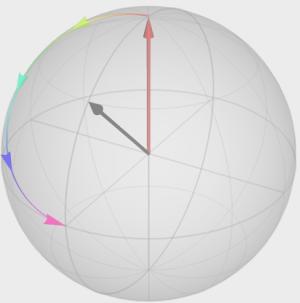
# Entropy

## Shannon entropy and readout error



# Bonus content

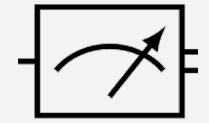
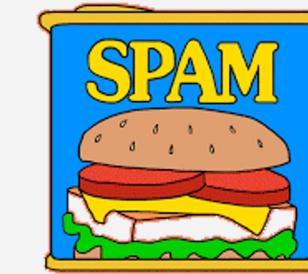
## Coherent noise



## Measurement in a nutshell Projection noise

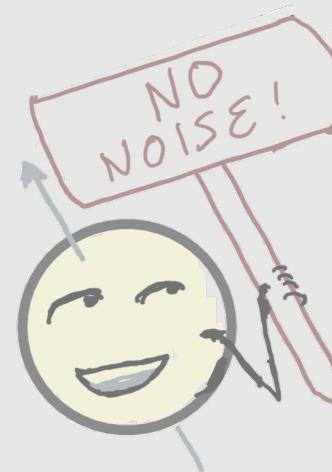
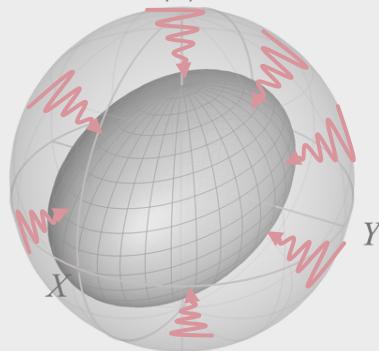
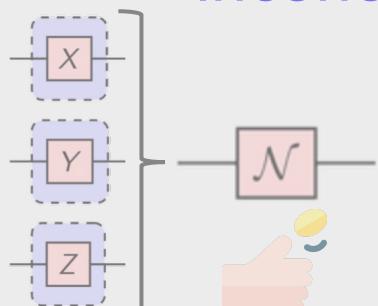


## SPAM: Noisy meters



$|0\rangle$  —

## Incoherent noise



## Bonus content



coin toss: flaticon; spam: make it move;  
road based on: freepik

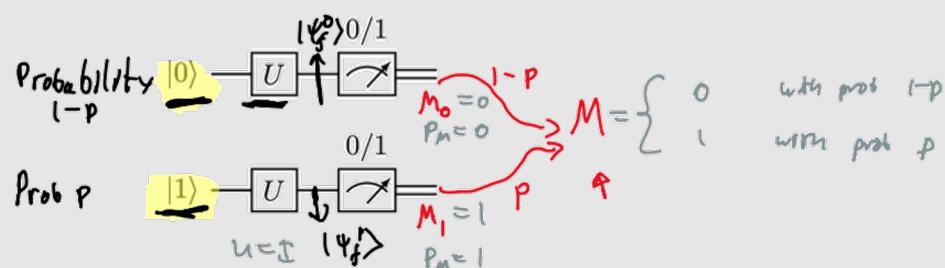
# State prep noise

## Introduction to quantum noise

### State preparation noise

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### Density operator

$$\rho = (1-p) |0\rangle \langle 0| + p |1\rangle \langle 1|$$

$\xrightarrow{\text{Pno error } \rho_{\text{ideal}}}$        $\xrightarrow{\text{Per err } \rho_{\text{error}}}$

$$\begin{aligned} \rho_f &= \text{Pno error } \rho_{\text{ideal}}^{\text{final}} + \text{Per err } \rho_{\text{error}}^{\text{final}} \\ &= (1-p) |\psi_f^0\rangle \langle \psi_f^0| + p |\psi_f^1\rangle \langle \psi_f^1| \\ &= (1-p) U |0\rangle \langle 0| U^\dagger + p U |1\rangle \langle 1| U^\dagger \\ &= U \left[ (1-p) |0\rangle \langle 0| + p |1\rangle \langle 1| \right] U^\dagger \end{aligned}$$

$$\rho_v = U_p \rho U_v^\dagger$$

$$\rho_v = U_p \rho U_v^\dagger$$

$$|0\rangle \langle 0| = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$|1\rangle \langle 1| = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

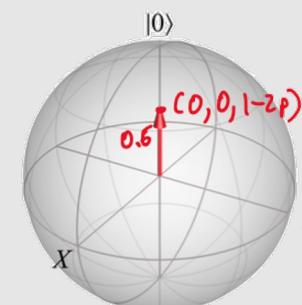
$$\rho = \begin{pmatrix} 1-p & 0 & 0 \\ 0 & 1-p & 0 \\ 0 & 0 & p \end{pmatrix} = \frac{1}{2} (I + (1-2p) \frac{1}{2} I)$$

$$\langle X \rangle = \text{Tr}[\Sigma X_\beta]$$

$$\langle Y \rangle = \text{Tr}[\Sigma Y_\beta]$$

$$\langle Z \rangle = \text{Tr}[\Sigma Z_\beta]$$

$$\text{Tr}[\rho_a \rho_b] = 2 \delta_{ab} \quad \text{for } a, b \in \{X, Y, Z\}$$



$$\begin{aligned} \rho &= \begin{pmatrix} 1 & 0 \\ 0 & 1-p \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 0 \end{pmatrix} \times 1-p \\ &\quad + \begin{pmatrix} 0 \\ 1 \end{pmatrix} \times p \end{aligned}$$

$$\text{Tr}(\rho^2) = \langle X \rangle^2 + \langle Y \rangle^2 + \langle Z \rangle^2 = 0^2 + 0^2 + (1-2p)^2$$



# State prep noise

Scaling to larger number of qubits

$$[(1-p)|0\rangle\langle 0| + p|1\rangle\langle 1|]^{\otimes n} = \boxed{U} \equiv \text{CNOT}$$

$$\rho_0 = [(1-p)|0\rangle\langle 0| + p|1\rangle\langle 1|]^{\otimes n}$$

$$= (1-p)^n |000\dots 0\rangle\langle 0\dots 0| + \dots |0100\dots\rangle\langle 1\dots 0| \dots$$

$$\Pr(M=00000\dots 0) = \begin{cases} 1 & \text{if } p=0 \text{ (ideal case)} \\ (1-p)^n \approx 1-np+O(p^2) & \text{for } p>0 \end{cases}$$

$$\langle Z_k \rangle = \langle \dots | Z | \dots \rangle = 1-2p \xrightarrow{\text{much faster}}$$

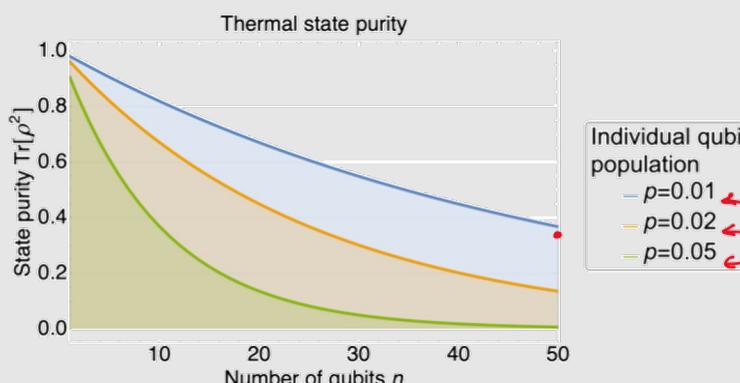
$$\langle ZZZ\dots Z \rangle = \text{Tr}_r [Z^{\otimes n} \rho] = (1-2p)^n$$

$$\rho_0 = \prod_{k=1}^n \frac{1}{2} (\hat{I}_k + (1-2p)\hat{Z}_k)$$

$$= \frac{1}{2^n} (1-\underline{z}) \otimes (1-\underline{z}) \otimes \dots \dots$$

$$= \frac{1}{2^n} \left[ (1-2p)^n \underline{Z}^{\otimes n} + \dots \dots \right]$$

$$\text{Tr}[\rho^2] = \prod_{k=1}^n \text{Tr}(\rho_{0,k}^2) = (1-2p)^{2n}$$



# The End!