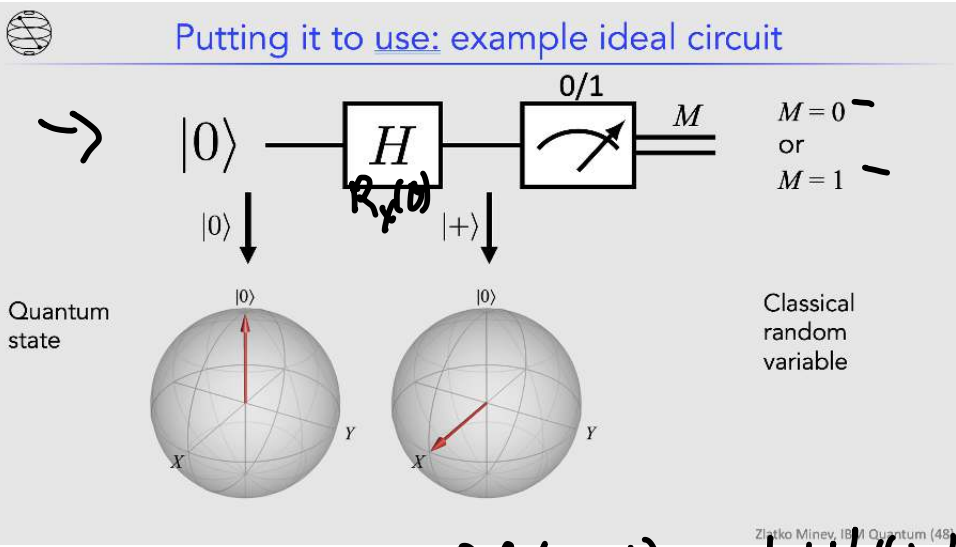
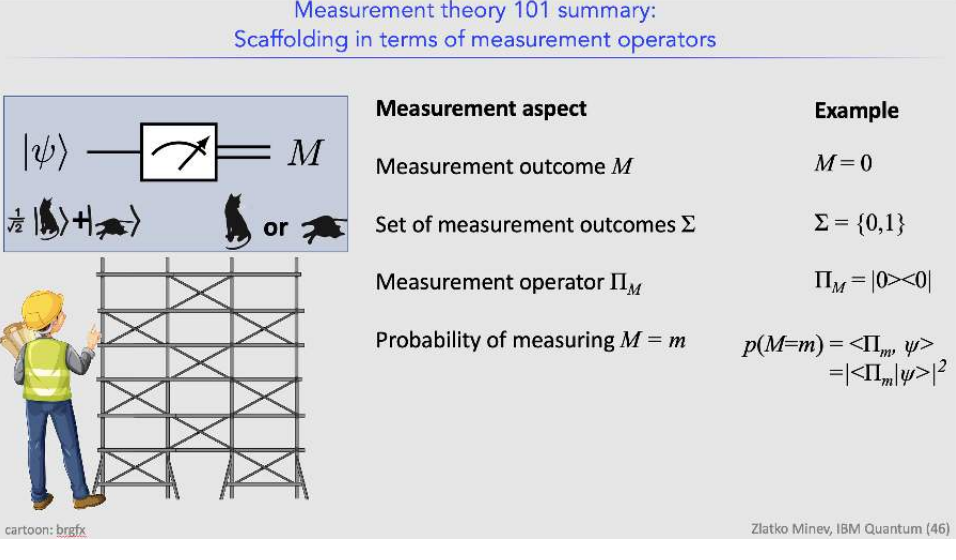
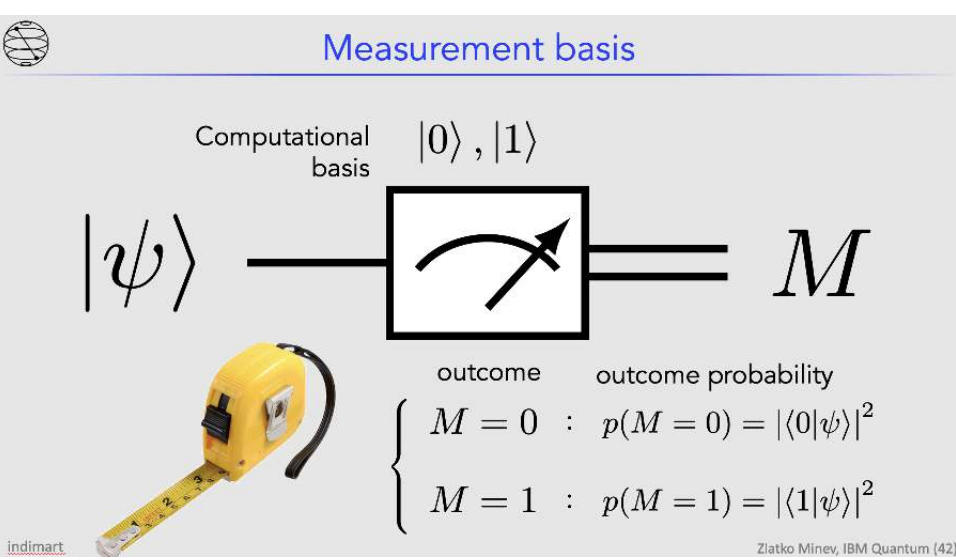


Introduction to quantum noise
Measurement theory & projection noise

Qiskit Global Summer School on Quantum Simulations

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$\hat{\pi}_0 = |0\rangle\langle 0|$
 $\hat{\pi}_1 = |1\rangle\langle 1|$
 $|\psi\rangle = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$
 $|\psi\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$
 $|\psi\rangle^\dagger = \begin{pmatrix} 1 & 0 \end{pmatrix}$

$p(M=1) = |\langle 1|\psi\rangle|^2$
 $p(M=0) = |\langle 0|\psi\rangle|^2$
 $= \langle \psi | \hat{\pi}_1 | \psi \rangle = \hat{\pi}_0 | \psi \rangle$
 $= \frac{1}{2} \begin{pmatrix} 1 & 1 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix}$
 $= \frac{1}{2} \cdot 1 = \frac{1}{2}$

$p(M=1) + p(M=0) = 1 \Rightarrow p(M=1) = 1 - p(M=0)$
 $M: \begin{cases} 1 & p=1/2 \\ 0 & p=1/2 \end{cases}$
Prob $\begin{matrix} 1/2 & 1/2 \\ \uparrow & \uparrow \\ M=0 & M=1 \end{matrix}$

Statistics: Mean
Random variable M drawn from distribution
 $\mathbb{E}[M] = \sum_{m \in \Sigma} m p(M=m)$
classical expectation value
 $= 0 \cdot p(M=0) + 1 \cdot p(M=1) = 0 \cdot \frac{1}{2} + 1 \cdot \frac{1}{2} = \frac{1}{2}$

more general case

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

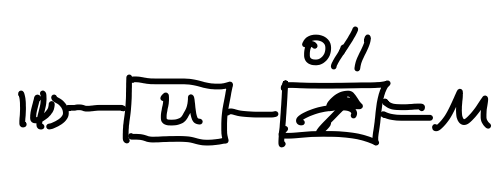
every possible qubit case

Bernoulli Distribution

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

\uparrow classical random variable
 \uparrow probability

Prob $\begin{matrix} 1-p & p \\ \uparrow & \uparrow \\ M=0 & M=1 \end{matrix}$



$\mathbb{E}[M] = \sum_{m \in \Sigma} m p(M=m)$ classical
 $= 0 p(M=0) + 1 p(M=1)$
 $= 0(1-p) + 1 \cdot p$
 $= p$

$\mathbb{E}[M] = \sum_{m \in \Sigma} m p(M=m)$ quantum
 $= \sum_{m \in \Sigma} m \langle \hat{\pi}_m \rangle$
 $= \sum_{m \in \Sigma} m \langle \psi | \hat{\pi}_m | \psi \rangle$
 $= \langle \psi | \sum_{m \in \Sigma} m \hat{\pi}_m | \psi \rangle$
 $= \langle \psi | \hat{M} | \psi \rangle$
 $= \langle \hat{M} \rangle$ observable

$\langle \hat{\pi}_m \rangle = \langle \psi | \hat{\pi}_m | \psi \rangle = \langle \psi | m \rangle \langle m | \psi \rangle$
linear
 $\langle a \hat{B} \rangle = a \langle \hat{B} \rangle$
 $\langle \hat{A} + \hat{B} \rangle = \langle \hat{A} \rangle + \langle \hat{B} \rangle$
define
 $\hat{M} := \sum_{m \in \Sigma} m \hat{\pi}_m$ defn

Different observables
For example $\hat{M} = 0|0\rangle\langle 0| + 1|1\rangle\langle 1| = \frac{1}{2}(\hat{\sigma}_z + \hat{I}) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$
 $\hat{M} = \frac{1}{2}(|0\rangle\langle 0| - |1\rangle\langle 1|) = \hat{Z} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$
for $\hat{M} = \hat{\sigma}_z$ $\Sigma = \{-1, +1\}$ Pauli observables

$$\hat{M} = \hat{X}$$

$$\mathcal{S} = \{+, -\}$$

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

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

$$\hat{X} = \sum_{m \in \mathcal{S}} m |m\rangle\langle m|$$

$$= +|+\rangle\langle +| - |-\rangle\langle -|$$

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

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

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

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

Statistic: Variance

$$V[M] = E[M^2] - E[M]^2$$

$$= \langle \hat{M}^2 \rangle - \langle \hat{M} \rangle^2$$

Return to $\mathcal{S} = \{0, 1\}$

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

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

$$p(M=1) = p$$

$$E[M^2] = \sum_{m \in \mathcal{S}} m^2 p(M=m)$$

$$= \sum m^2 \langle |m\rangle\langle m| \rangle$$

$$= \langle \sum m^2 |m\rangle\langle m| \rangle$$

$$= \langle \hat{M}^2 \rangle$$

$$= \langle |1\rangle\langle 1| \rangle$$

$$= \langle \hat{\Pi}_1 \rangle$$

$$= p = E[M]$$

$$V[M] = p - p^2$$

$$= p(1-p)$$

$$= \sigma_M^2$$

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

$$\langle \hat{M}^2 \rangle = \langle \psi | \hat{M} \cdot \hat{M} \cdot | \psi \rangle$$

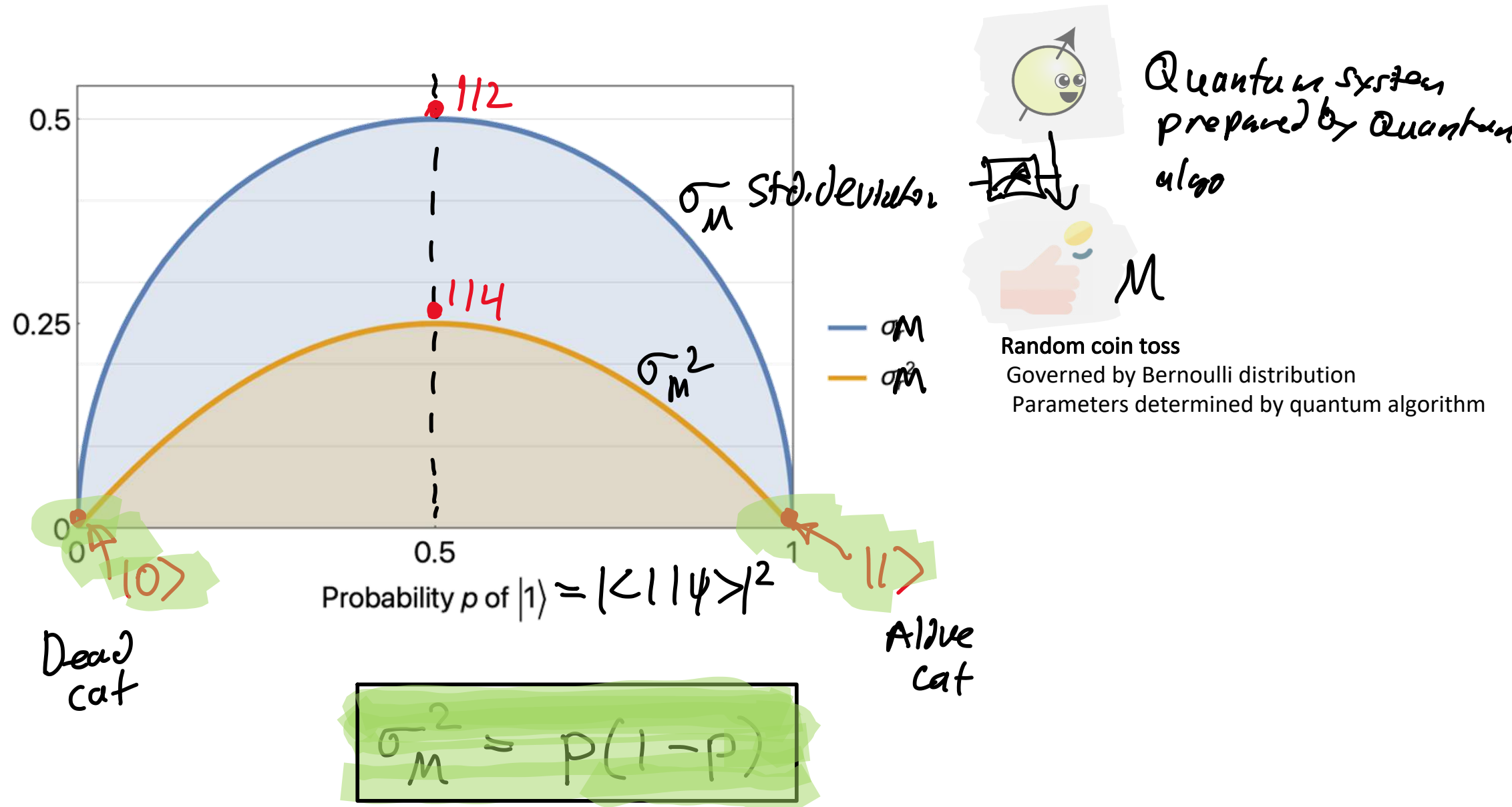
$$p(M=m) = \langle |m\rangle\langle m| \rangle$$

$$\hat{M} = 0|0\rangle\langle 0| + 1|1\rangle\langle 1|$$

$$\hat{M}^2 = 1|1\rangle\langle 1| = M$$

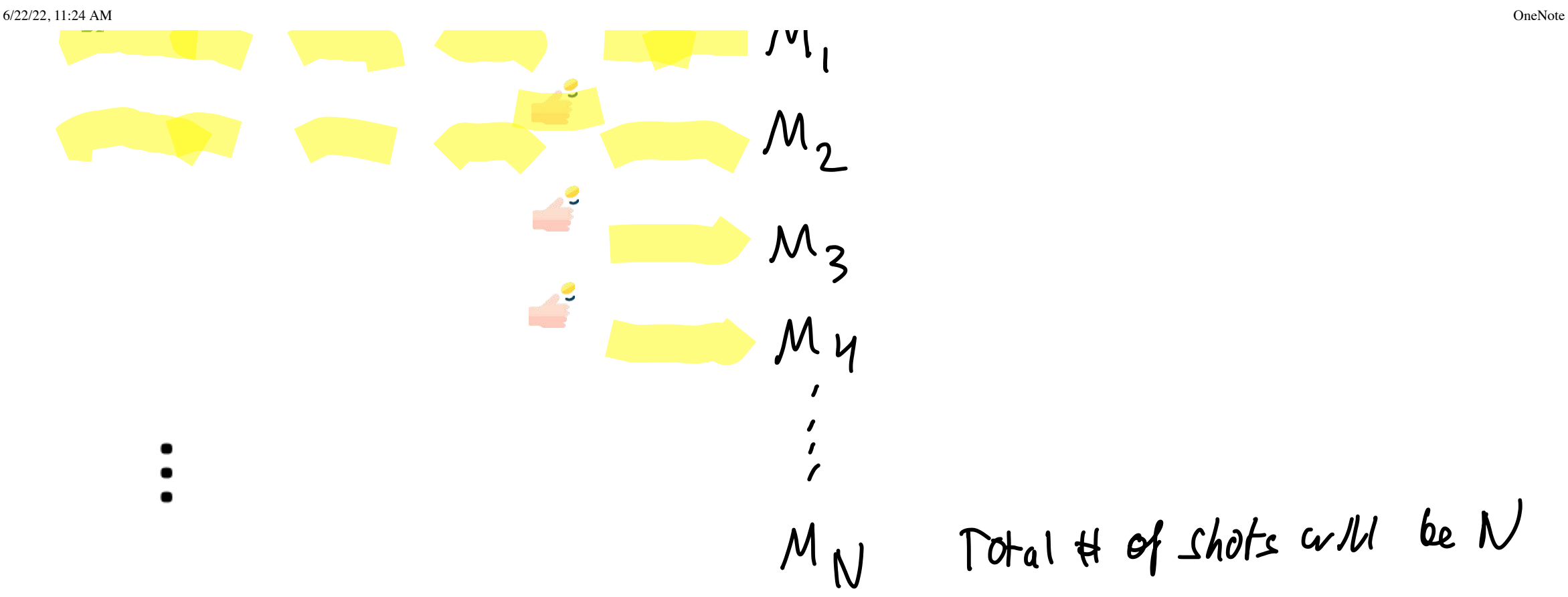
0 iff $p=0$ or $p=1$

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



Projection noise and sampling error

Let's turn to the example of finite number of shots we execute for our experiment.
Perform N experiment, each giving us a single shot result 0 or 1



For 3 samples, there are 2^3 possible outcome sequences ($N=3$)

$$\begin{matrix}
 m_1 & m_2 & m_3 \\
 0 & 0 & 1 \\
 0 & 0 & 0 \\
 0 & 1 & 1 \\
 1 & 0 & 0 \\
 1 & 0 & 1 \\
 1 & 1 & 0 \\
 1 & 1 & 1
 \end{matrix}$$

Sample mean?

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

\nwarrow random variable
 \nwarrow # shot
 \nwarrow index of the shot #
 \nwarrow indep. identically distributed (i.i.d)
 $M_n \sim M$

2^N possible combinations \rightarrow examp $S = \frac{1}{3}(1+0+1) = 2/3$

$$\begin{aligned}
 \mathbb{E}[S] &= \mathbb{E}\left[\frac{1}{N} \sum_{n=1}^N M_n\right] \\
 &= \frac{1}{N} \sum_{n=1}^N \mathbb{E}[M_n] \quad \mathbb{E}[M_n] = \mathbb{E}[M_1] = \mathbb{E}[M] \\
 &= \frac{1}{N} \mathbb{E}[M] \\
 &= \mathbb{E}[M]
 \end{aligned}$$

$$\mathbb{E}[S] = \mathbb{E}[M] = \langle \hat{M} \rangle = p \text{ for prob } \hat{M} = (1 \times 1)$$

$$V[S] = \frac{V[M]}{N} = \frac{p(1-p)}{N}$$

$\sigma_S = \sqrt{\frac{p(1-p)}{N}}$
 error bar (std dev)
 on sample mean
 estimator for our expectation value.